

# Signal and Imaging Sciences Workshop 1999 Proceedings

*J. V. Candy*

**January 5, 2001**

***U.S. Department of Energy***

Lawrence  
Livermore  
National  
Laboratory



Signal and Imaging Sciences

# Workshop

1999

# PROCEEDINGS

CENTER FOR ADVANCED SIGNAL AND IMAGING SCIENCES





## DISCLAIMER

This document was prepared as an account of work sponsored by an agency of the United States Government. Neither the United States Government nor the University of California nor any of their employees, makes any warranty, express or implied, or assumes any legal liability or responsibility for the accuracy, completeness, or usefulness of any information, apparatus, product, or process disclosed, or represents that its use would not infringe privately owned rights. Reference herein to any specific commercial product, process, or service by trade name, trademark, manufacturer, or otherwise, does not necessarily constitute or imply its endorsement, recommendation, or favoring by the United States Government or the University of California. The views and opinions of authors expressed herein do not necessarily state or reflect those of the United States Government or the University of California, and shall not be used for advertising or product endorsement purposes.

This report has been reproduced  
directly from the best available copy.

Available to DOE and DOE contractors from the  
Office of Scientific and Technical Information  
P.O. Box 62, Oak Ridge, TN 37831  
Prices available from (615) 576-8401, FTS 626-8401

Available to the public from the  
National Technical Information Service  
U.S. Department of Commerce  
5285 Port Royal Rd.,  
Springfield, VA 22161



# Table of Contents

## NOVEMBER 11 OPENING PRESENTATIONS

A Microphone Array for Hearing Aids, <i>Dr. Bernard Widrow - Keynote Speaker:</i> .....	6
---	---

## SPEECH PROCESSING

Background Speaker Noise Removal Using Combined EM Sensor / Acoustic Signals, <i>Larry Ng</i> .....	20
Speaker Verification Performance Comparison Based on Traditional and EM Sensor Pitch Extraction, <i>Todd Gable</i> .....	30
Defining a Human Voiced Excitation Function using Glottal Electromagnetic Micro Power Sensors (GEMS), <i>Greg Burnett</i> .....	38
Synapse TAP™ Universal Access Technology, <i>Martin Tibor</i> .....	48

## COMPUTED TOMOGRAPHY

Confocal Ultrasound Imaging through Acoustically Thick Media, <i>Sean Lehman</i> .....	50
Filtered Backprojection for Low Contrast Medical Ultrasound Tomography Using Matlab, <i>David Scott</i> .....	61
Optimal Beam Hardening Correction for X-Ray Tomography, <i>Jesse Kolman</i> .....	66
Performance Evaluation of Amorphous-Silicon Flat Panel Array for X-Ray Digital Radiography and Computed Tomography, <i>Ken Dolan</i> .....	74

## ASTRONOMICAL IMAGING

Speckle Imaging of Saturn's Moon Titan, <i>Seran Gibbard</i> .....	86
Astronomical Imaging Fourier Transform Spectroscopy: Technology and Techniques, <i>Ron Wurtz</i> .....	90
"First Light" Science Results From the Keck Telescope Adaptive Optics System, <i>Claire Max</i> .....	99
Searches for Low Mass Extra-solar Planets, <i>Jian Ge</i> .....	108
Demonstration of a Diffractive Telescope, <i>Sham Dixit</i> .....	117

## IMAGE PROCESSING

Detection of Vertical Obstructions in SAR Images, <i>Sailes Sengupta</i> .....	126
Use of Morphological Operators and Pattern Recognition Techniques for Sorting the FIRST Data, <i>Deanne Proctor</i> .....	127
Non-invasive Recovery of Local Properties of Deforming Objects from Registered Range Data, <i>Leonid Tsap</i> .....	142
Subsurface Gap Depth Detection by Infrared Imaging Using a Surface Heat Pulse, <i>Charles Landram</i> .....	151



## NOVEMBER 12 OPENING PRESENTATIONS

Overview of Engineering, <i>Spiros Dimolitsas – Associate Director, Engineering</i> .....	154
A Retrospective on Computer Vision Research, <i>Dr. Avi Kak, Keynote Speaker</i> .....	158
Overview of Biopatheogen Detection Technology at LLNL, <i>Fred Milanovich</i> .....	159

## PROCESSING FOR COMPLEX SYSTEMS

Overview of the Center for Complex Distributed Systems, <i>Dave McCallen</i> .....	162
Using Signal Processing Techniques to Improve Finite Element Modeling, <i>Greg Burnett</i> .....	163
Damage Detection for Enhanced Evaluation of Structures, <i>Greg Clark</i> .....	171
Parabolic Reflector for Radar Mine Detection, <i>Steve Azevedo</i> .....	181

## SIGNAL PROCESSING

Broad Band Acoustic Ranging and Velocimetry, <i>Dave Chambers</i> .....	191
Grating Lobe Reduction in Large Element, 2-Dimensional Phased Arrays, <i>Jan-Ulco Kluiwstra</i> .....	200
Denoising Data using Wavelet Based Methods, <i>Chuck Baldwin</i> .....	201
Hematoma Detection Using MIR, <i>David Scott</i> .....	219

## IMAGING AND DETECTION

Towards Automated Optical Evaluation of Protein Crystals, <i>Pat Fitch</i> .....	228
In-Situ Sensing of Nucleation and Crystal Growth of $\text{Cd}_{1-x}\text{Zn}_x\text{Te}$ , <i>William Choi</i> .....	229
Phase Recovery in X-ray Diffraction, <i>Abraham Szoke</i> .....	240

## SIGNAL/IMAGE PROCESSING FOR NDE

Nondestructive Characterization Center Overview, <i>Harry Martz</i> .....	249
Overview of NDE Activities at LLNL, <i>Graham Thomas</i> .....	255
Signal Processing for Laser Ultrasonic NDE, <i>Robert Huber</i> .....	265
Matched Field Imaging for NDE, <i>James Candy</i> .....	273
Void Analysis from CT Imagery with Applications to Damage Evolution in AM60B Magnesium Alloy, <i>Amy Waters</i> .....	283
Ultrasonic Techniques for Materials Characterization and Process Monitoring, <i>Peter Martin</i> .....	297
Gamma-ray Imaging with a Segmented Germanium Detector, <i>Greg Schmid</i> .....	303



---

Day 1

Opening

**Presentations**

---



# A Microphone Array for Hearing Aids

*Bernard Widrow*  
*Prof. of E.E., Stanford*

A directional acoustic receiving system is constructed in the form of a necklace including an array of two or more microphones mounted on a housing supported on the chest of a user by a conducting loop encircling the user's neck.

Signal processing electronics contained in the same housing receives and combines the microphone signals in such a manner as to provide an amplified output signal which emphasizes sounds of interest arriving in a direction forward of the user. The amplified output signal drives the supporting conducting loop to produce a representative magnetic field. An electroacoustic transducer including a magnetic field pick-up coil for receiving the magnetic field is mounted in or on the user's ear and generates an acoustic signal representative of the sounds of interest.

The microphone output signals are weighted (scaled) and combined to achieve desired spatial directivity responses. The weighting coefficients are determined by an optimization process. By bandpass filtering the weighted microphone signals with a set of filters covering the audio frequency range and summing the filtered signals, a receiving microphone array with a small aperture size is caused to have a directivity pattern that is essentially uniform over frequency in two or three dimensions. This method enables the design of highly-directive hearing instruments, which are comfortable, inconspicuous, and convenient to use. The invention provides the user with a dramatic improvement in speech perception over existing hearing aid designs, particularly in the presence of background noise and reverberation.



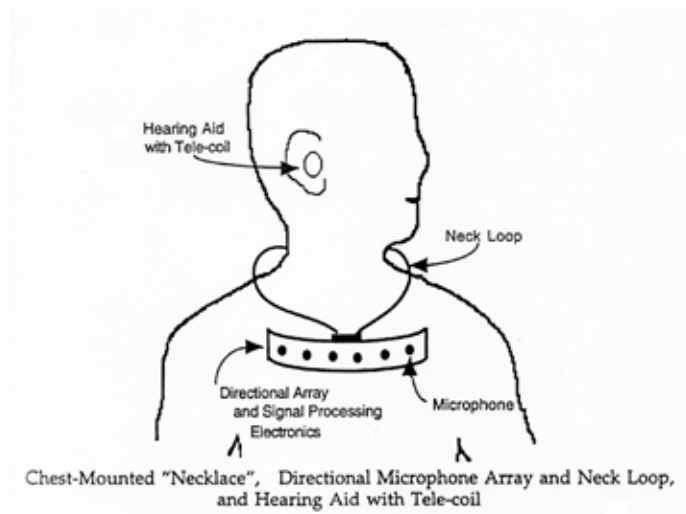
Cardinal Sound Labs

## The Array

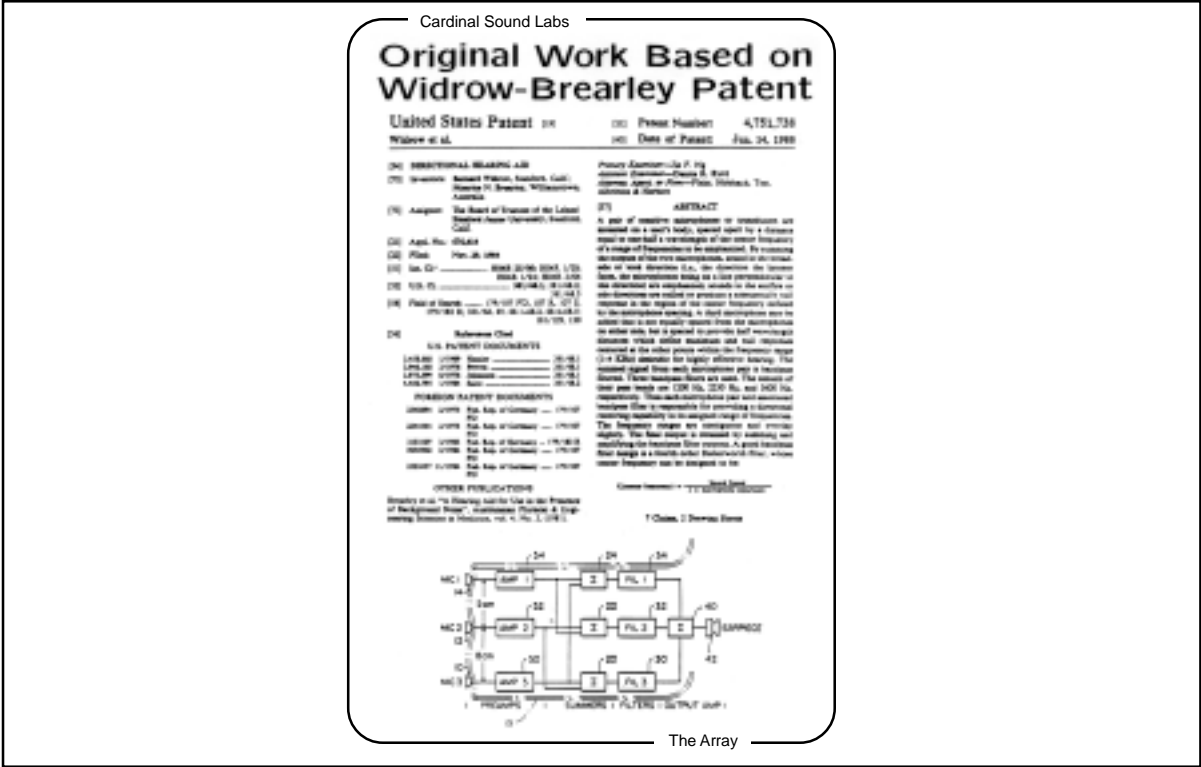
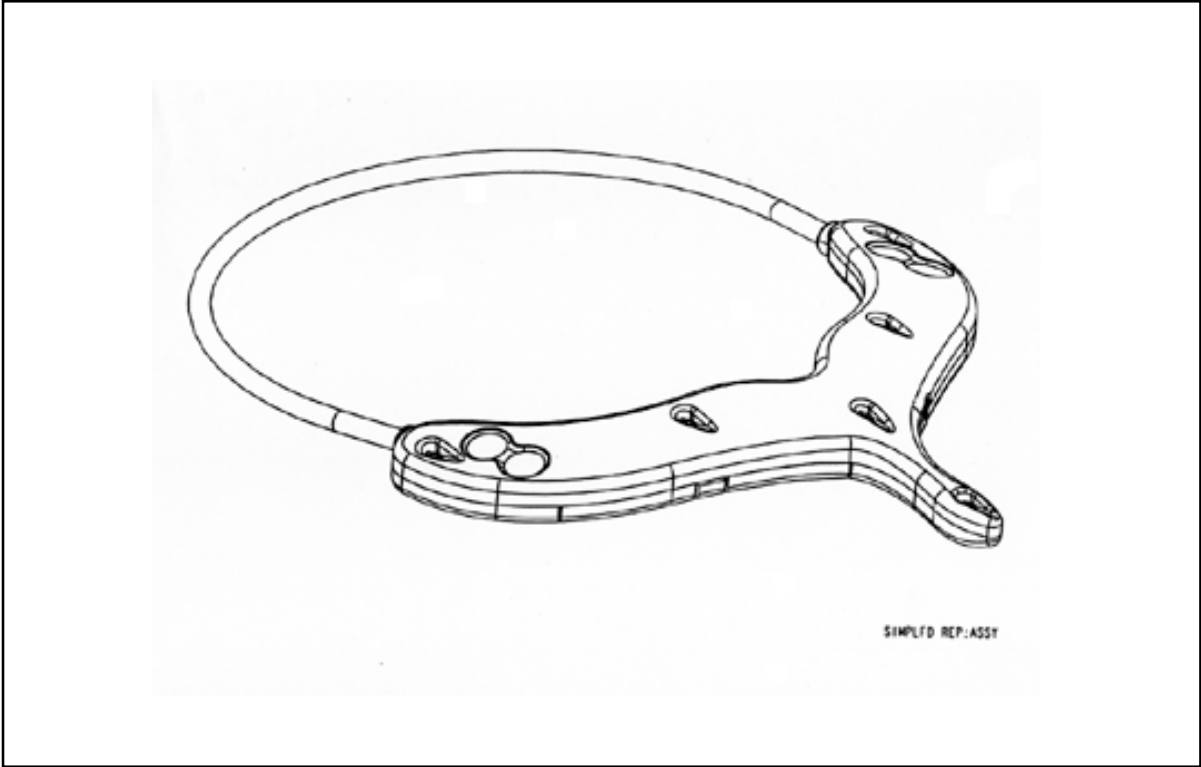
Bernard Widrow  
Chairman  
Cardinal Sound Labs

The Array

Page 1



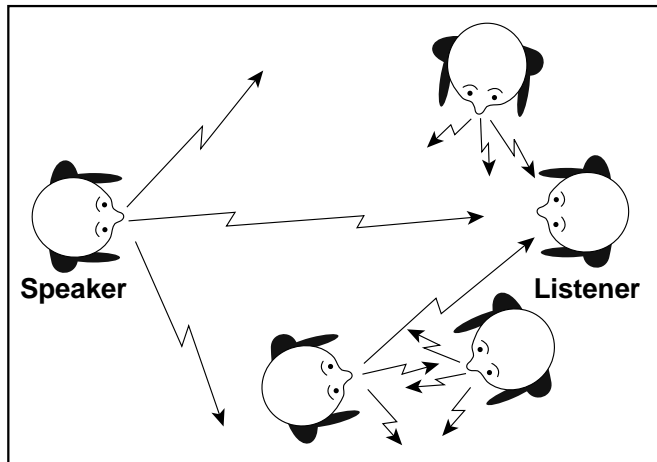






Cardinal Sound Labs

## Speech Discrimination Suffers in the Presence of Noise

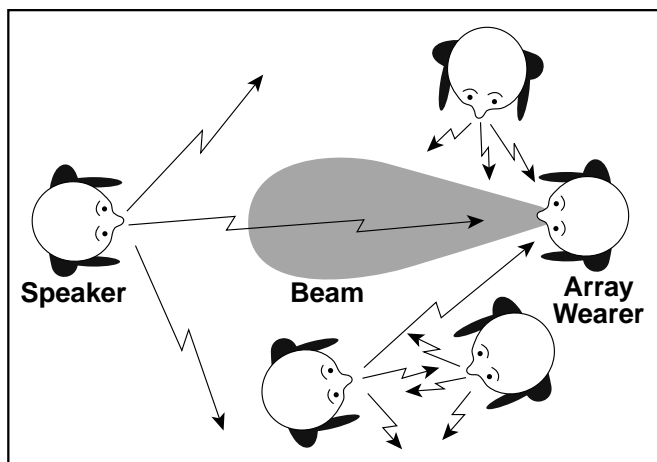


The Array

Page 3

Cardinal Sound Labs

## The Array's Beam Rejects Sound from Extraneous Sources



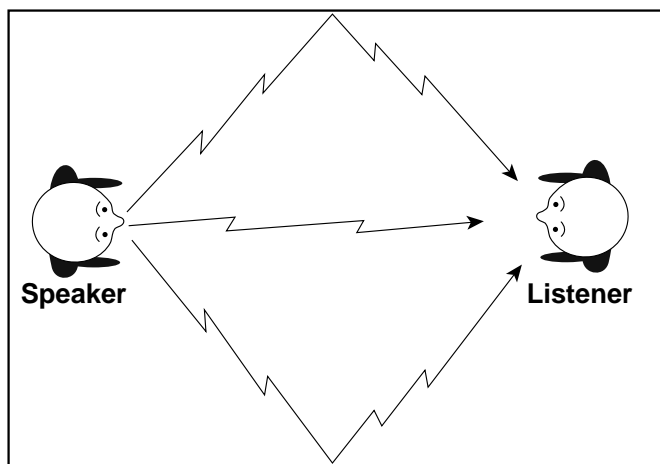
The Array

Page 4



Cardinal Sound Labs

## In a Reverberant Room, Echoes Interfere with Speech Discrimination

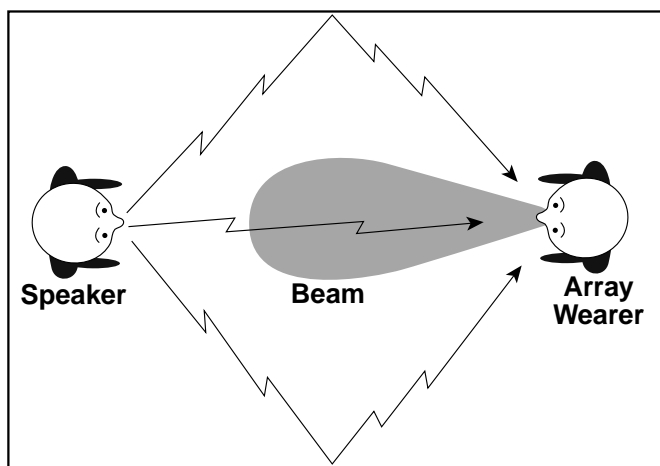


The Array

Page 5

Cardinal Sound Labs

## The Array's Beam Rejects Echoes, Improving Speech Discrimination



The Array

Page 6



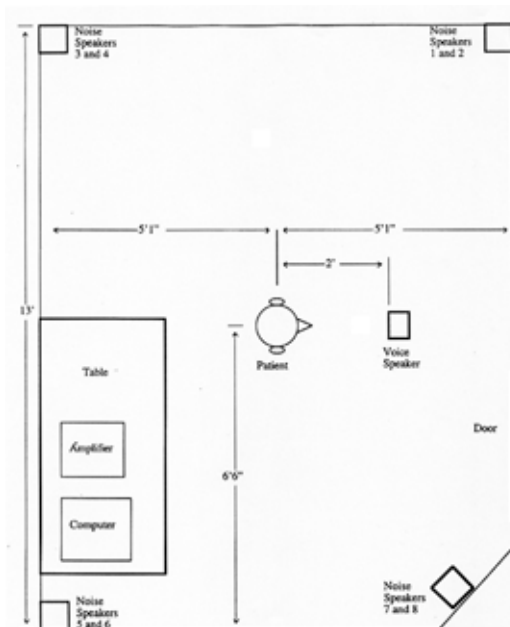
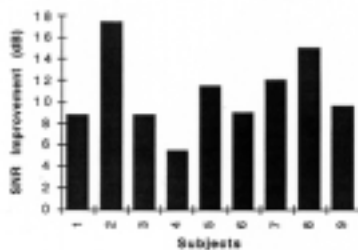
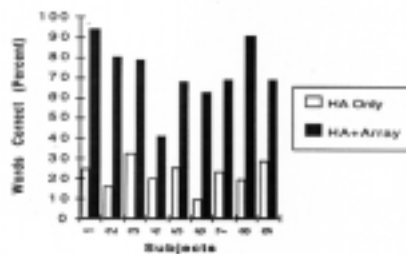


Fig. 3 Arrangement of test room at GSAA.

Page 11



a. Magnitude of improvement in sentence speech recognition threshold in noise (HENT) with the microphone array, in comparison with the hearing aid alone.



b. Percent speech intelligibility in the presence of noise. For each test subject, the yellow bar is the result obtained using the hearing aid and the solid bar is the result using the array.

Fig. 4. Results obtained at GSAA on HENT and intelligibility test.

Page 12



Cardinal Sound Labs

## Patch Cord Enables Direct Link to TV, Sony Walkman, and other Devices

- **Sony Walkman**
- **Television**
- **Telephone**
- **Airline Movies**
- **Corded Microphone for Automobile**
- **Infrered Hookup for Theaters**



The Array

Page 9

Cardinal Sound Labs

## The Array Has Several Other Useful Characteristics

- **Can be worn and operated under clothing or outside clothing.**
- **Wireless connection to hearing aid improves comfort and convenience of system.**
- **System greatly reduces squealing problem of hearing aids by reducing acoustic feedback.**
- **Battery life of final design expected to exceed one month.**

The Array

Page 10



Cardinal Sound Labs

## Companding and AGC Provides Comfort and Very Low Circuit Noise

- **Compander brings SNR of Array output down to level of Knowles microphones.**
- **Compander uses a 60-decibel range in 120 1/2 decibel steps** on front and back end.
- **Array has infinity-to-one AGC with user adjustable sensitivity** (kneepoint) control.
- **Sensitivity adjustment allows feedback and low-level sounds** to be removed without affecting nearby sounds.
- **Companding, AGC, and sensitivity control also applied to** external audio input.

The Array

Page 11

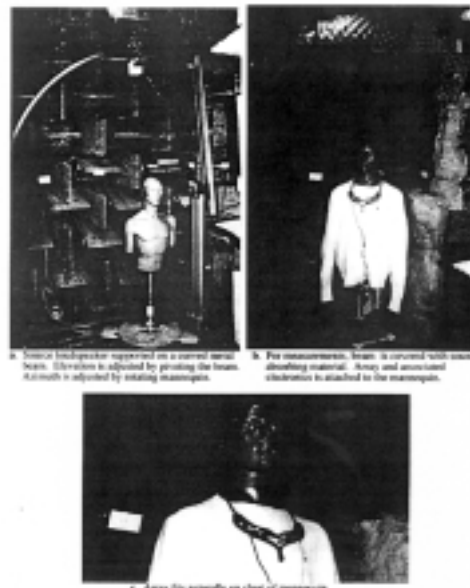
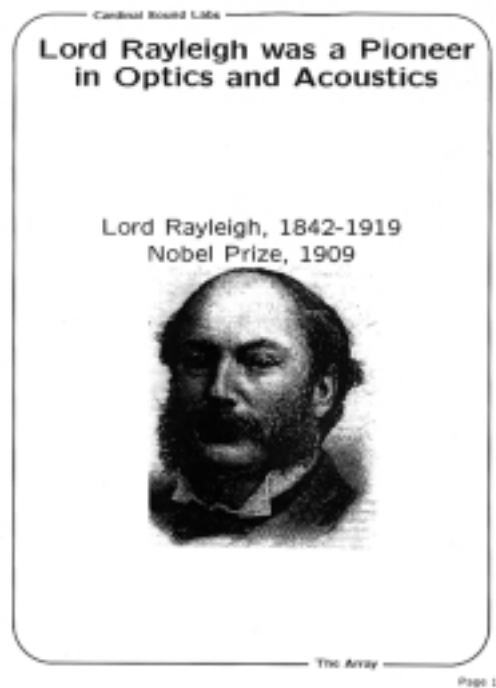
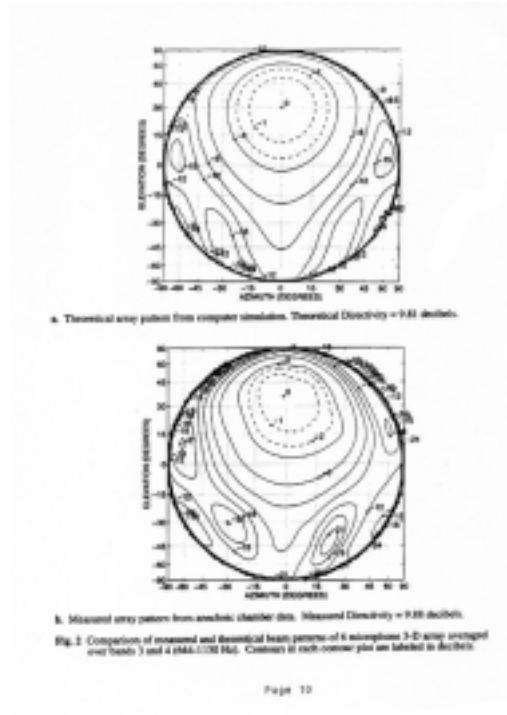


Fig. 1 Testing the 6-microphone 3-D array in an anechoic chamber.

Page 9







Cardinal Sound Labs

## Lord Rayleigh's Formula Predicts Minimum Size for an Array

Lord Rayleigh's formula is:

$$\left( \begin{array}{c} \text{BEAMWIDTH} \\ \text{IN RADIANS} \end{array} \right) = \frac{1}{\text{APERTURE SIZE}}$$

This formula tells us that to get a beamwidth of 60° (approximately 1 radian) at frequencies as low as 500 Hertz, the width and height of the Cardinal Sound Labs' Array would need to be approximately two feet.

The Array

Page 13

Cardinal Sound Labs

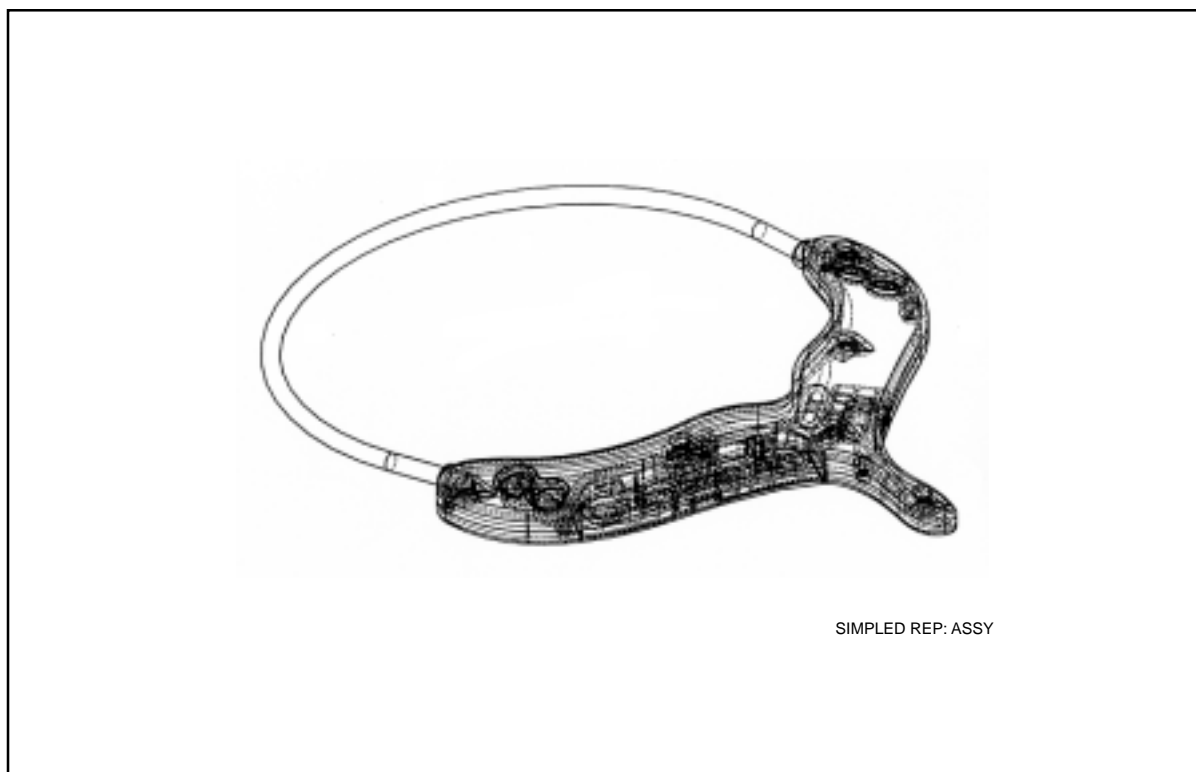
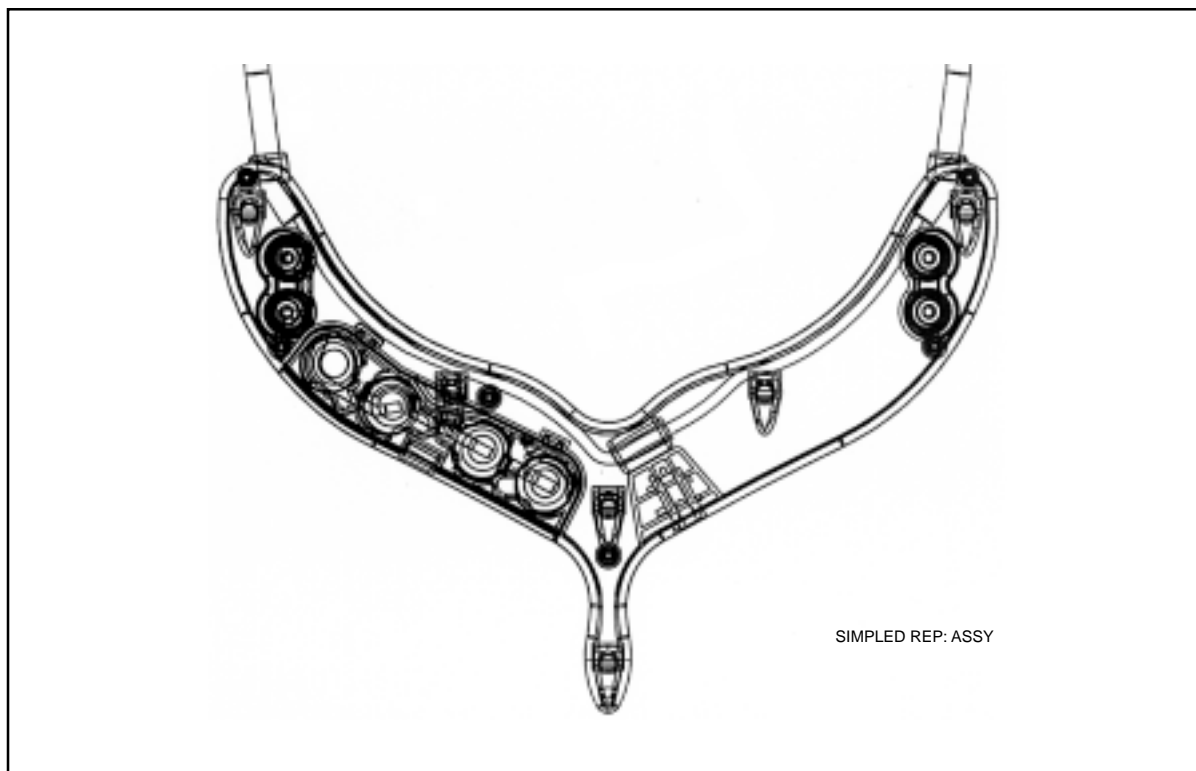
## New Prototypes Get Past Size Limitation

We have overcome this limitation without violating the laws of physics, and this is the substance of our most recent patent application. The Array is approximately 7.7 inches wide and 5 inches high.

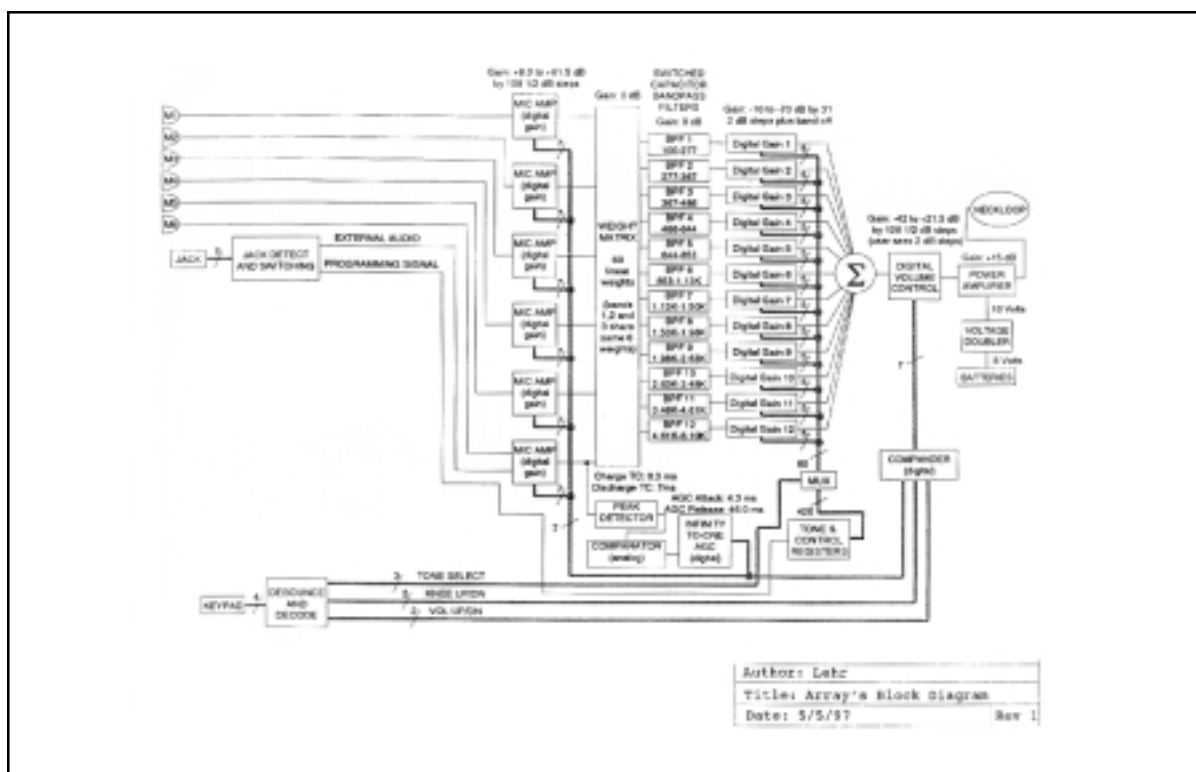
The Array

Page 14















---

# Speech Processing

---



# Background Speaker Noise Removal Using Combined EM Sensor/Acoustic Signals

*L.C. Ng, G. C. Burnett, J. F. Holzrichter, and T. J. Gable*

Recently, very low power EM radar-like sensors have been used to measure the macro- and micro-motions of human speech articulators as human speech is produced [ see Holzrichter et. al. JASA. 103 (1) 622, (1998)]. These sensors can measure tracheal wall motions, associated with the air pressure build up and fall as the vocal folds open and close, leading to a voiced speech excitation function. In addition, they provide generalized motion measurements of vocal tract articulator gestures that lead to speech formation. For example, tongue, jaw, lips, velum, and pharynx motions have been measured as speech is produced. Since the EM sensor information is independent of acoustic air pressure waves, it is independent of the state of the acoustic background noise spectrum surrounding the speaker. By correlating the two streams of information together, from a microphone and (one or more) EM sensor signals, to characterize a speaker's speech signal, much of the background speaker noise can be eliminated in real-time. This paper presents the improved performance of glottal EM sensors (GEMS) based algorithms in suppressing a variety of background speaker noise. [Work supported by NSF and DOE.]

## Background Speaker Noise Removal Using Combined EM Sensor/Acoustic Signals

Signal and Imaging Sciences Workshop

Nov 11, 1999

Presented by:

**Lawrence C. Ng, Ph.D.**  
**Group Leader**

Signal/Image Processing and Control Group  
Defense Sciences Engineering Division

Collaborators:

Dr. J. Holzrichter, Dr. G. Burnette, Mr. T. Gable

Lawrence Livermore National Laboratory  
P.O. Box 808, L-491, Livermore, CA, 94551



## Overview



- Brief Description of GEMS (sensor) Characteristics
- Application to Speech Processing
- The Background Noise Removal Problem
- The Wiener Filter Solution
- The Glottal Windowing (GWIN) Filter Solution
- The Glottal Correlation (GCOR) Filter Solution
- Summary

MTP-991111-LCN-A - 2

## Micro Power EM Sensors (Radar-related) Were Developed as High Speed Diagnostic Tools for a Laser Fusion Experiment at LLNL



- Emits pulses of very low power ( $< 1 \text{ nW}$ ) and short duration
- Center frequency is around 2.5 GHz (but adjustable)
- Due to short duration of pulse it is composed of many different frequencies
- Emits pulses at a 2 MHz PRF which results in high signal to noise ratio (many returns can be averaged)
- Operating in a field disturbance mode, it is an excellent motion and vibration detector

Transmitter/Receiver Electronics

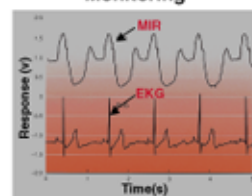


MTP-991111-LCN-A - 3

Motion Detector

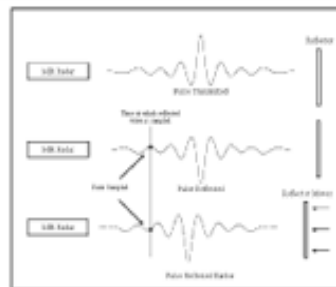


Heart Monitoring



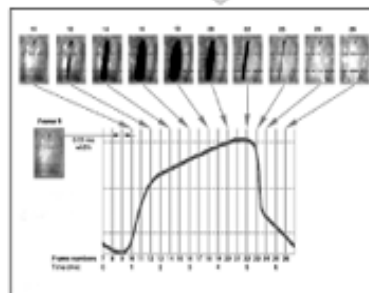
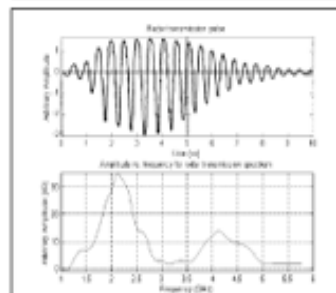


## EM Sensors Derive the Signal by Measuring The Differential Positions of the Returned Wave



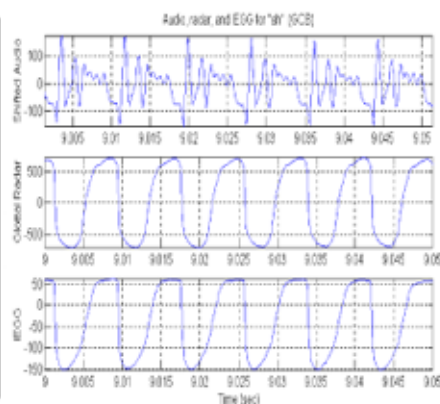
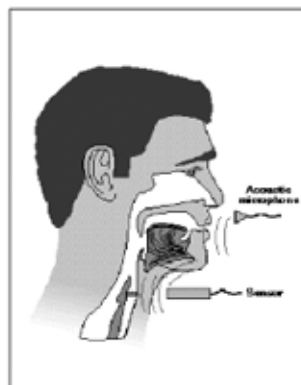
EM sensor measures the changes in signal amplitude resulting from the movement of an object

Good correlation of EM sensor signal to high speed video recording of vocal folds motion was obtained



MTP-991111-LCN-A - 4

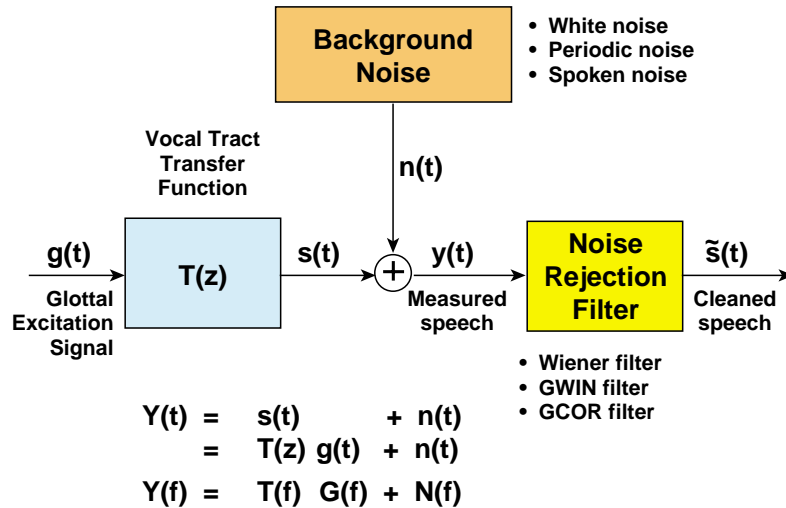
## Glottal EM Sensor (GEMS) Tissue Measurements Show Strong Correlation with Vocal-Fold Electrolglottography (EGG) Signal



MTP-991111-LCN-A - 5

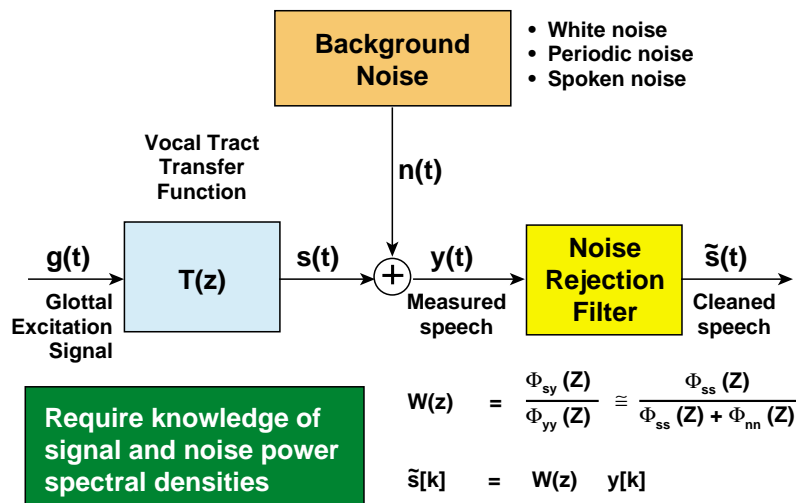


## Background Noise Removal Problem Must Address Two Major Issues: Voice detection and Noise Rejection



MTP-991111-LCN-A - 6

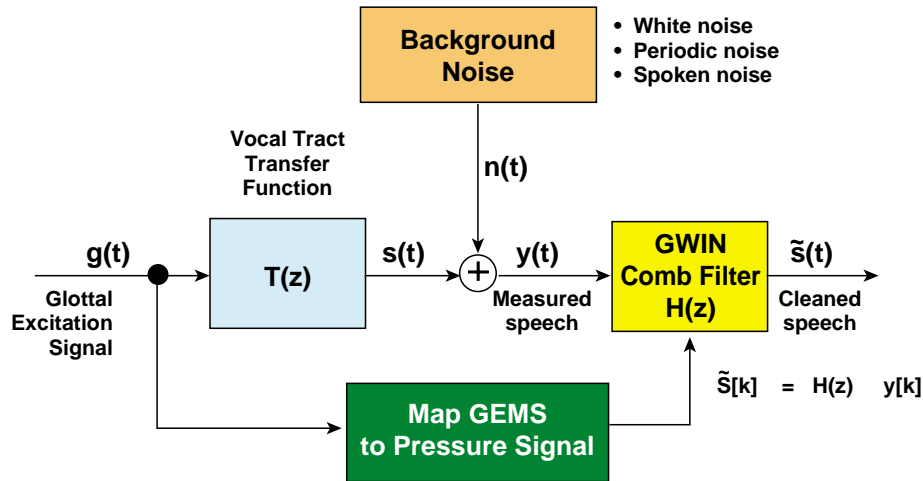
## The Wiener Filter Can Optimally Minimize the Noise When the Signal and Noise Spectrum Are Available



MTP-991111-LCN-A - 7

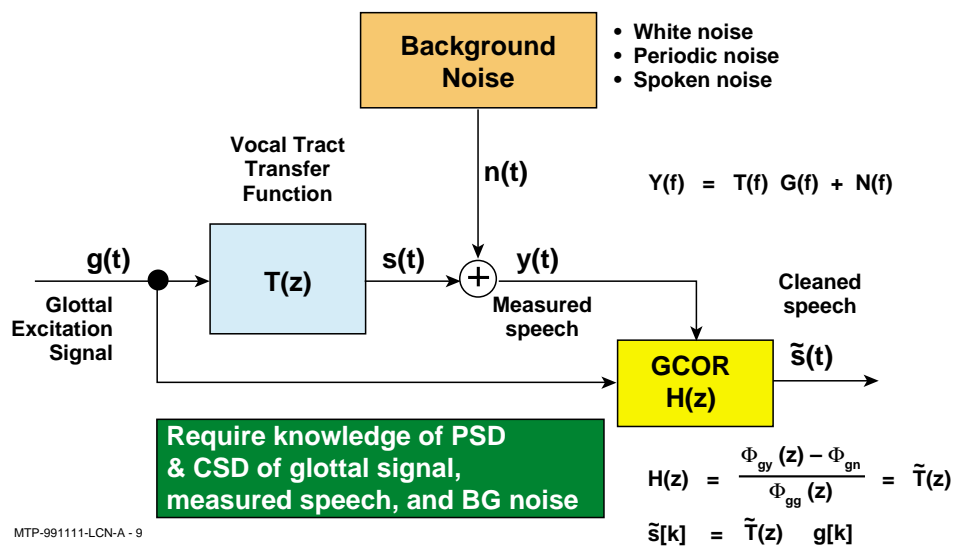


## The Glottal Windowing (GWIN) Filter Can Be Simply Constructed from the Glottal EM Sensor Signal



MTP-991111-LCN-A - 8

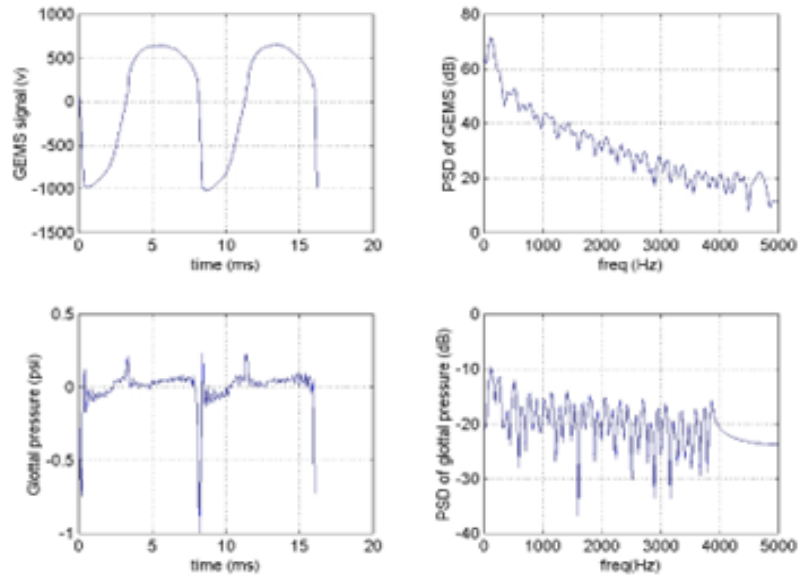
## Glottal Correlation (GCOR) Filter Can Be Used To Optimally Remove the Background Noise



MTP-991111-LCN-A - 9

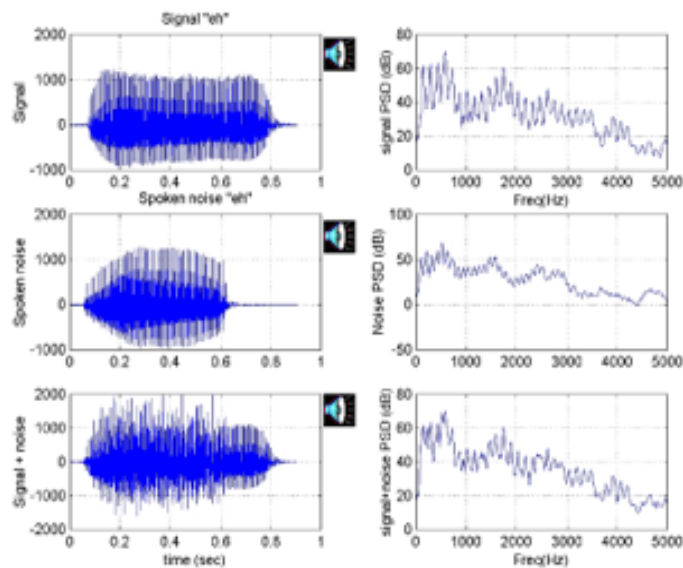


## GWIN Comb Filter Can Be Easily Constructed From Glottal EM Sensor Measurements



MTP-991111-LCN-A - 10

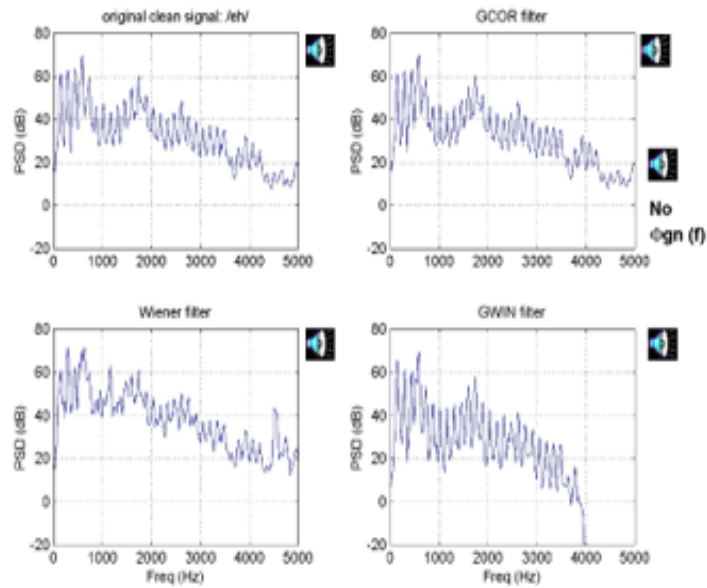
## We Added the Utterance /eh/ Spoken by Two Males Simultaneously as a Background Noise Test Problem



MTP-991115-LCN.A - 11

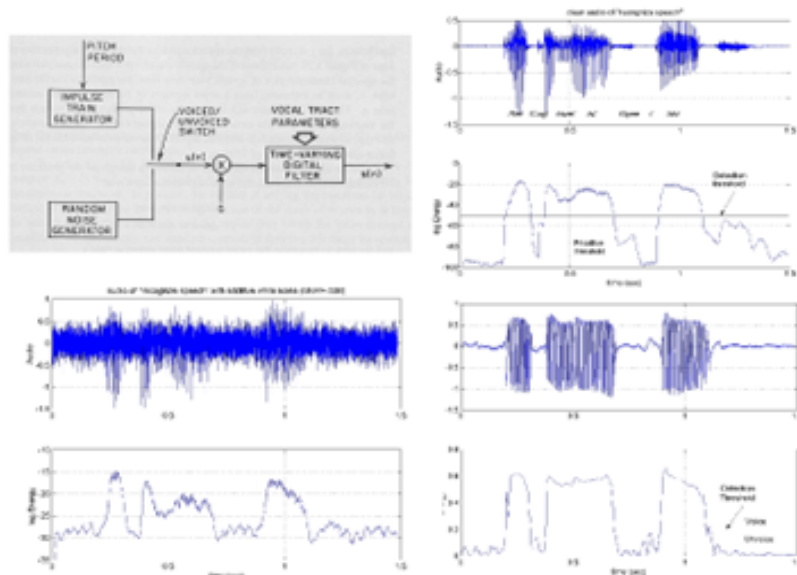


## We Observed That Both GWIN and GCOR (Glottal based) Performed Better Than the Wiener Filter



MTP-991111-LCN-A -12

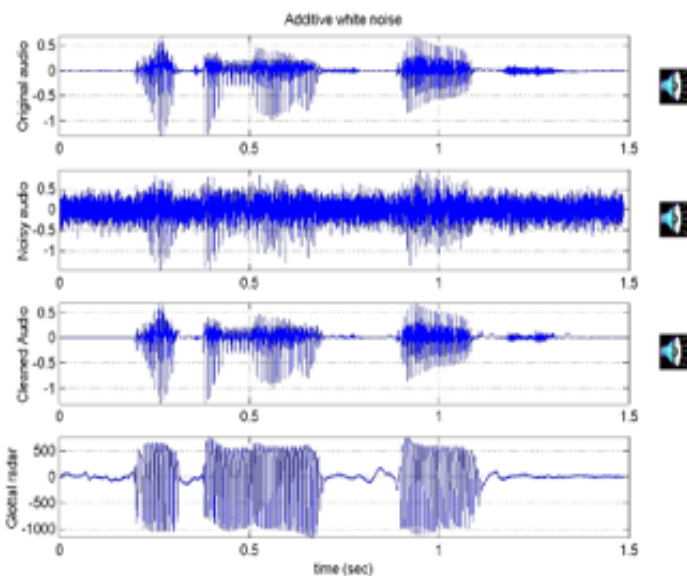
## Detection of Voiced and Unvoiced Boundary is Difficult in a High Background Noise Environment



MTP-991111-LCN-A -13

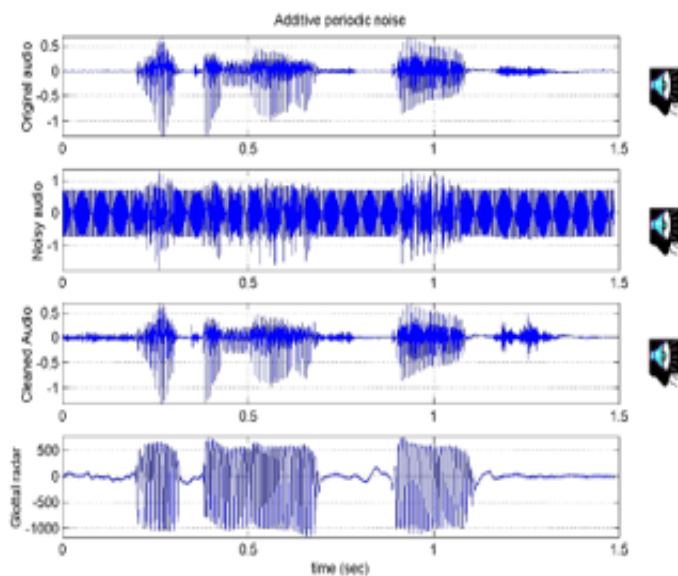


### Background White Noise Suppression with the Spoken Words: "Recognize Speech"



MTP-991111-LCN-A - 14

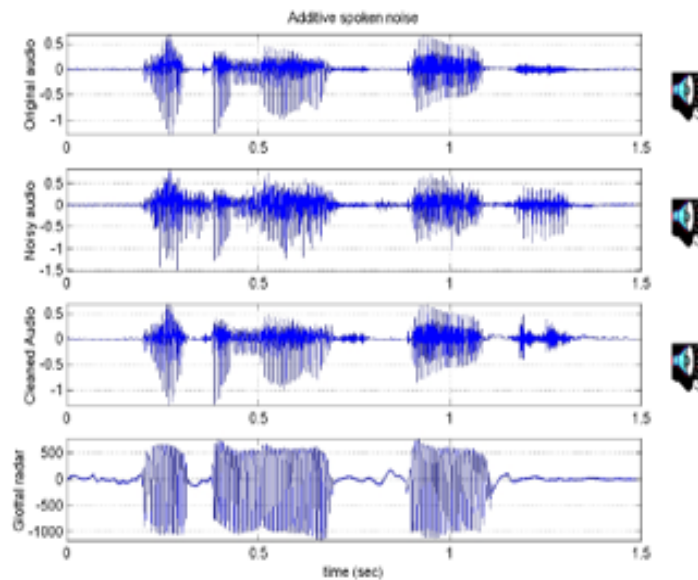
### Background Periodic Noise Suppression with the Spoken Words: "Recognize Speech"



MTP-991111-LCN-A - 15



### Background Spoken Noise Suppression with the Spoken Words: "Recognize Speech"



MTP-991111-LCN-A -16

### GEMS-Microphone - Could Provide a Reliable and Robust Voice Interface in a Noisy Environment



MTP-991111-LCN-A -17



## Summary and Conclusions



- We have examined several options to remove background noise
  - Wiener Filter (classical approach)
  - GWIN Filter (Spectral comb filter generated from pressure waveform derived from GEMS measurement)
  - GCOR Filter (make use of glottal and voice signal correlation)
- We tested the filters with three kinds of background noise
  - White noise
  - Periodic noise
  - Spoken noise
- Both GWIN and GCOR out performed the classical Wiener filter
  - Signal and noise spectrum are difficult to obtain for transient speech signal
  - GEMS is useful in detecting voiced/unvoiced speech boundaries
- Measured GEMS glottal signal is a key to remove background noise
- We applied GEMS based filters to various background noise and found that GEMS is also effective against spoken noise
- GEMS-microphone has a potential to operate in a very noisy environment

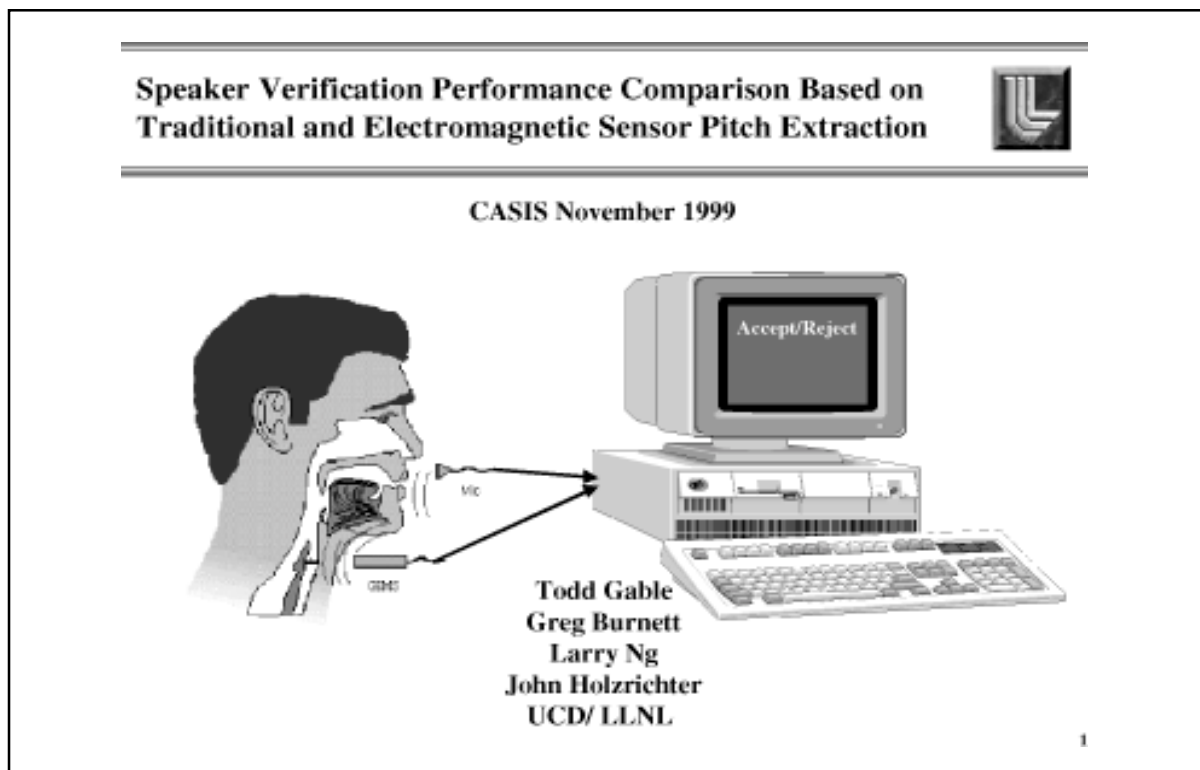
MTP-991111-LCN-A -18



## Speaker Verification Performance Comparison Based on Traditional and Electromagnetic Sensor Pitch Extraction

*T.J. Gable, L.C. Ng, G.C. Burnett, and J.F. Holzrichter  
Lawrence Livermore National Laboratory*

This work compares the speaker verification performance between a traditional acoustic-only pitch extraction to a new electromagnetic (EM) sensor based pitch approach. The pitch estimation approach was developed at the Lawrence Livermore National Laboratory (LLNL) utilizing Glottal Electromagnetic Micropower Sensors (GEMS, also see <http://speech.llnl.gov/>). This work expands previous pitch detection work by Burnett et. al. (to be published IEEE Trans. on Speech and Audio Processing) to the specific application of speaker verification using dynamic time warping. Clearly, a distinct advantage of GEMS is its insensitivity to acoustic ambient noise. This work demonstrates the clear advantage of the GEMS pitch extraction to improve speaker verification error rates. Cases with added white noise and other speech noise were also examined to show the strengths of the GEMS sensor in these conditions. The EM sensor speaker verification process operated without change over signal-to-noise (SNR) conditions ranging from -20dB to -2.5dB, the acoustic algorithms became unusable at SNR exceeding -10db. [Work supported by NSF and DOE.]





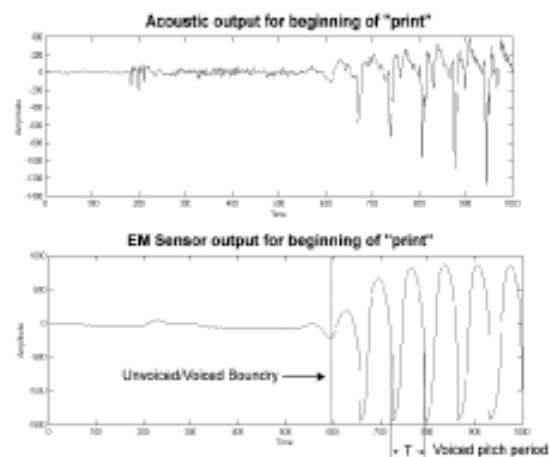
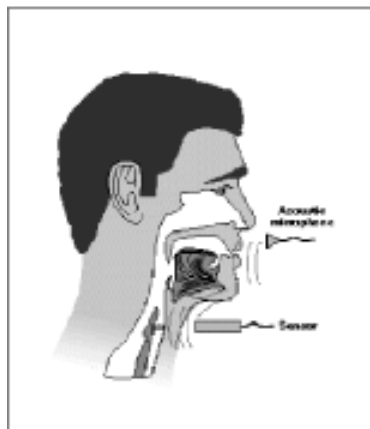
**Purpose of this study is to evaluate the performance of EM sensor (GEMS) using pitch in speaker verification**



- Comparing performance differences between traditional acoustic and Glottal Electromagnetic Sensor (GEMS) pitch extraction using a corpus of TIMIT sentences from 15 male speakers
- Using GEMS extracted pitch and conventional dynamic time warping verification methods lead to lower speaker verification errors
- Demonstrate the GEM sensor's insensitivity to acoustic noise

2

**Glottal EM Sensor (GEMS) signal is well suited for voiced speech pitch extraction**



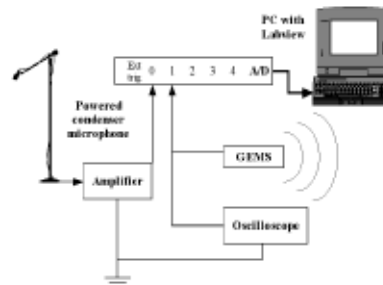
3



### GEMS and acoustic data was collected from 15 speakers



- PC Laptop with Labview and NiDaQ was used to take data
  - Data was sampled at 40kHz and filtered and decimated to 10kHz in post-processing
- 15 male volunteers were used
  - 12 sentences were recorded with 2 sets of 5 repetitions were recorded, separated by at least 2 weeks

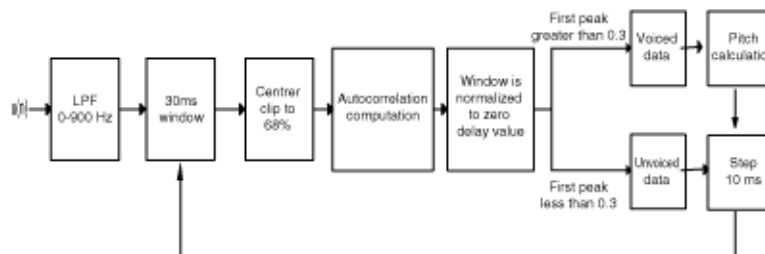


4

### Traditional acoustic autocorrelation pitch calculation



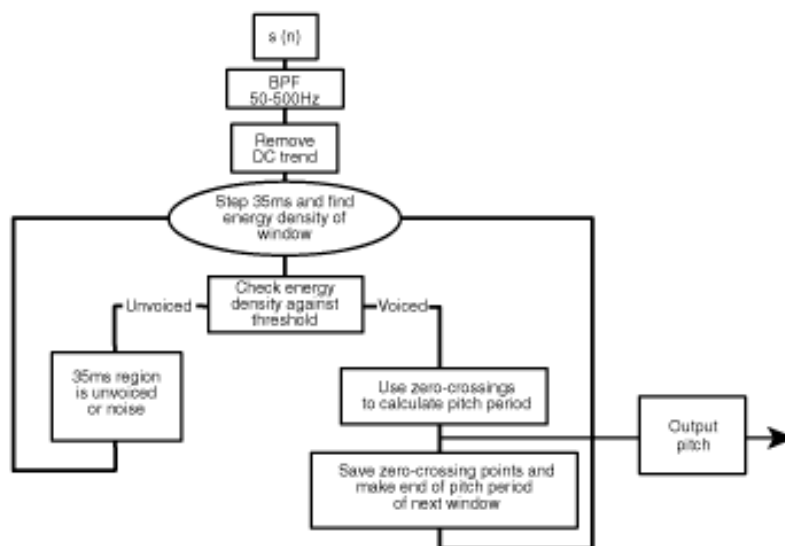
$$r_l(m) = \frac{1}{N-m} \sum_{n=0}^{N-m} [u(n+l)w(n)][x(n+l+m)w(n+m)]$$



5



### GEM sensor pitch extraction



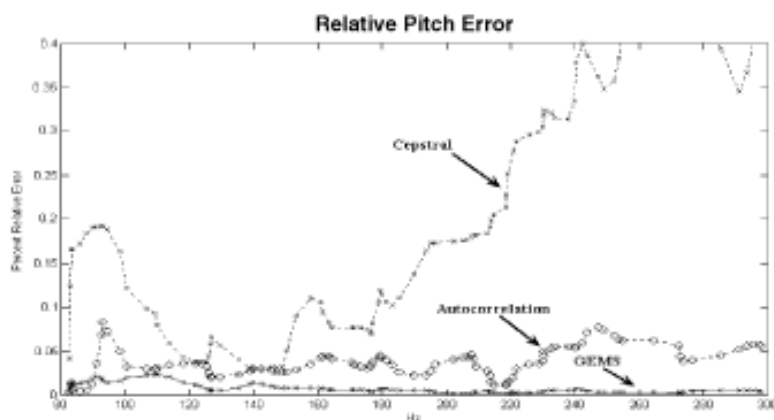
6

### Pitch extraction performance comparison



Pick  $f_k$  at random to four decimal places

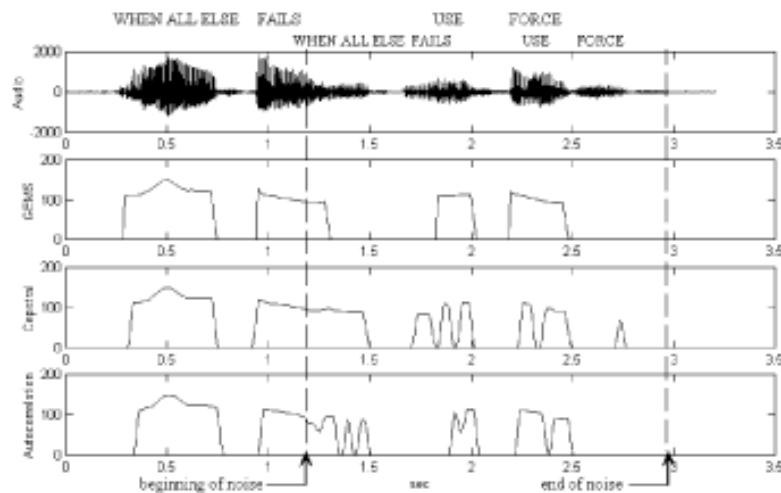
$$\text{Synthetic signal } s(t) = \sum_{N=1}^{20} \frac{1}{N} \sin(2\pi N f_k t)$$



7



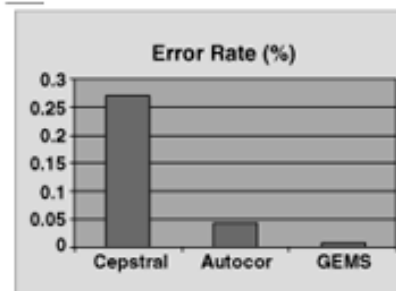
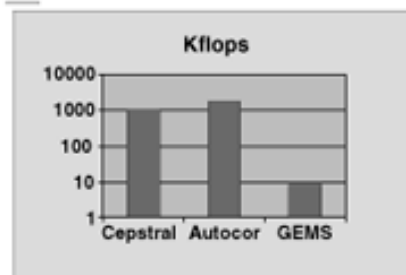
### Example of showing that GEMS' pitch estimate is unaffected by acoustic noise



### From Tuning Fork Experiments We Determine Pitch Estimate Performance of the GEMS



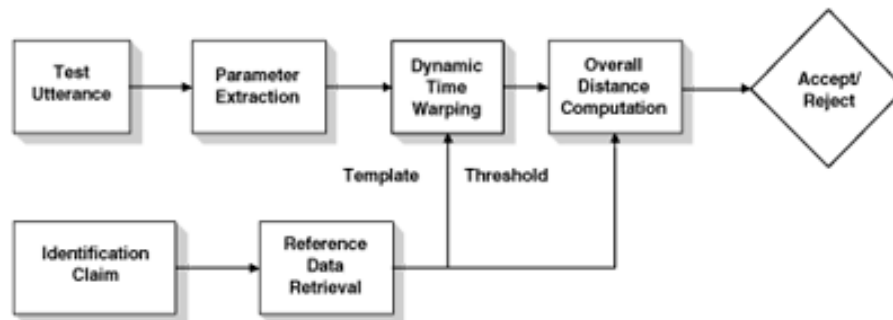
- Generate known signals from tuning fork experiments
  - Compute pitch estimate using cepstral, autocorr, and GEMS zero-crossing
- Results indicated that GEMS pitch estimate is:
  - 100 times faster in computational efficiency
  - 5 to 20 times smaller in error rates
- Conclusions
  - GEMS is insensitive to acoustic noise
  - Fast computation of pitch can provide real-time pitch synchronous speech processing
  - Performance improve with higher sampling rate



9



**Speaker verification is the process of accepting or rejecting a persons identity claim based on parameters extracted from a test utterance**

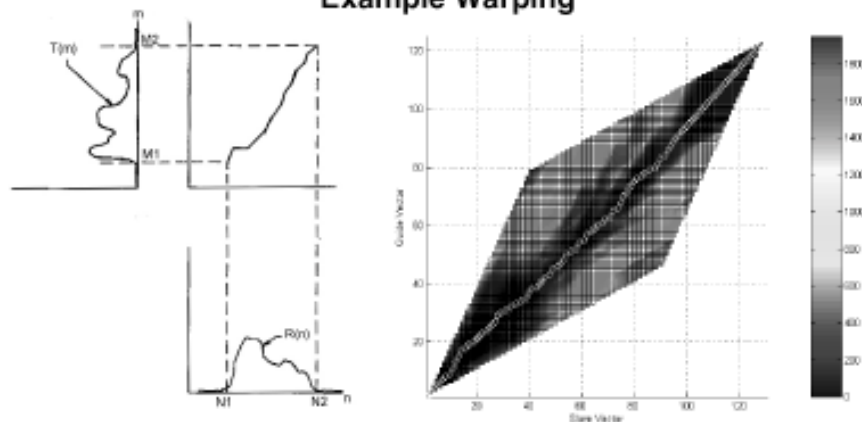


10

**Dynamic Time Warping (DTW) is the process of warping two vectors to the same length which yields a distance measure that mathematically tells us how similar they are**



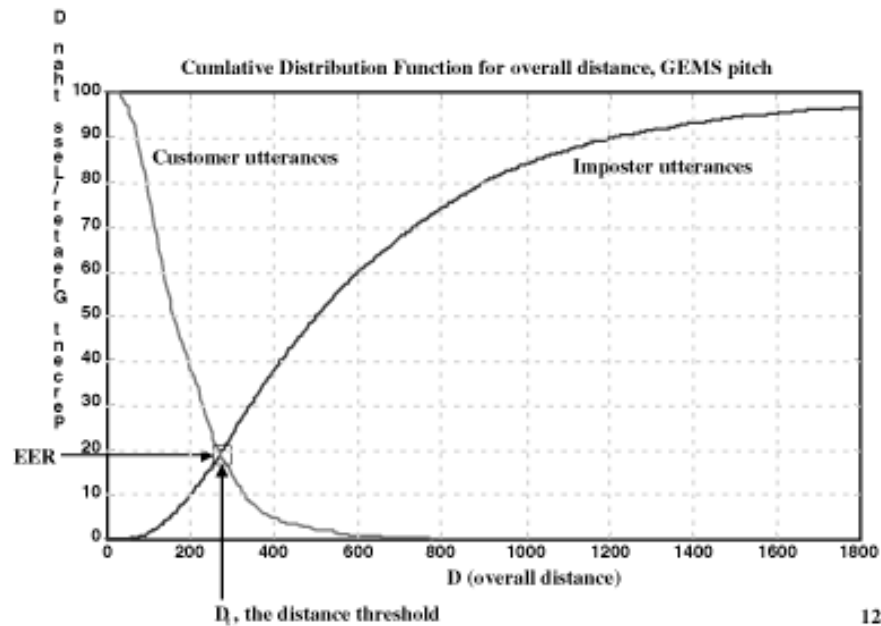
### Example Warping



11

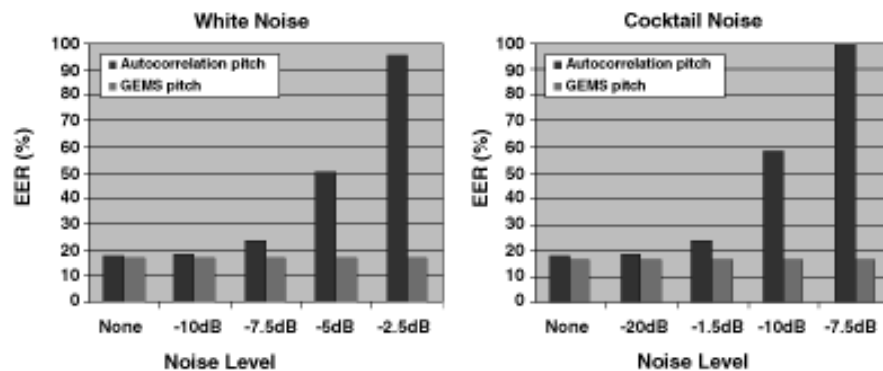


## How to find the Equal Error Rate (EER) from the warping distances



12

## GEMS equal error rate (EER) is insensitive to acoustic noise



13



## Conclusions

---



- The GEMS is a small non-invasive, low power, low cost device well suited for pitch extraction
- The GEMS sensor is fast, accurate and noise insensitive
- With this more accurate and robust pitch extraction method, pitch can be used more reliably in speaker verification
- With the present database, the GEMS extracted pitch based verification is about 10% better than acoustic autocorrelation pitch based verification. Additional EM sensor based features will further reduce verification error rates.

14



## Defining a Human Voiced Excitation Function using Glottal Electromagnetic Micropower Sensors (GEMS)

*T.J. Gable, L.C. Ng, G.C. Burnett, and J.F. Holzrichter  
Lawrence Livermore National Laboratory, POB 808, Livermore CA 94551*

This work compares the speaker verification performance between a traditional acoustic-only pitch extraction to a new electromagnetic (EM) sensor based pitch approach. The pitch estimation approach was developed at the Lawrence Livermore National Laboratory (LLNL) utilizing Glottal Electromagnetic Micropower Sensors (GEMS, also see <http://speech.llnl.gov/>). This work expands previous pitch detection work by Burnett et. al. (to be published IEEE Trans. on Speech and Audio Processing) to the specific application of speaker verification using dynamic time warping. Clearly, a distinct advantage of GEMS is its insensitivity to acoustic ambient noise. This work demonstrates the clear advantage of the GEMS pitch extraction to improve speaker verification error rates. Cases with added white noise and other speech noise were also examined to show the strengths of the GEMS sensor in these conditions. The EM sensor speaker verification process operated without change over signal-to-noise (SNR) conditions ranging from -20dB to -2.5dB, the acoustic algorithms became unusable at SNR exceeding -10db. [Work supported by NSF and DOE.]

## Defining a Human Voiced Excitation Function using Glottal Electromagnetic Micropower Sensors (GEMS)



**CASIS Signal and Image Sciences Workshop**

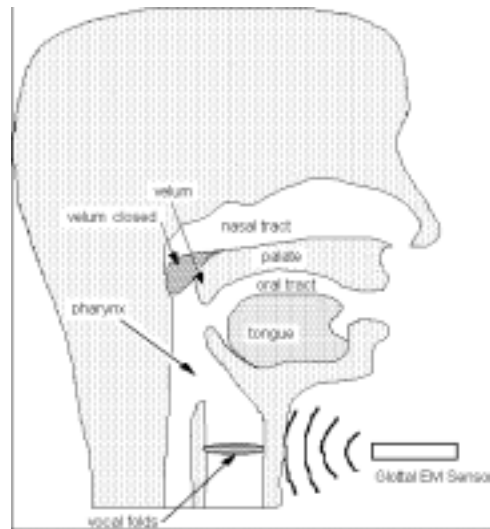
**November 11, 1999**

**Greg Burnett**

**Lawrence Livermore National Laboratory**

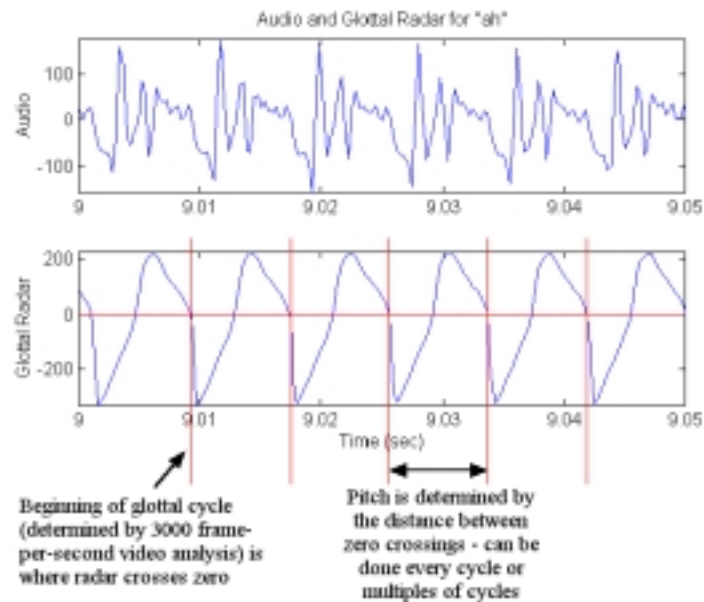


## Glottal EM sensor tissue measurements



- The GEMS can detect tissue motion all the way up and down the trachea, not just near the vocal folds
- This motion can be detected at skin to GEMS distances of up to 6 cm

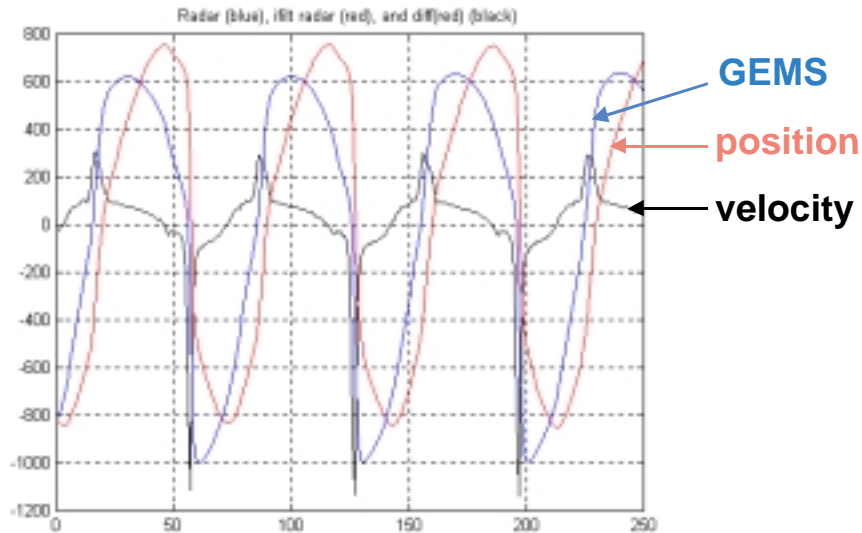
## Raw GEMS signal coming from neck



CASIS Nov. 99 p.3

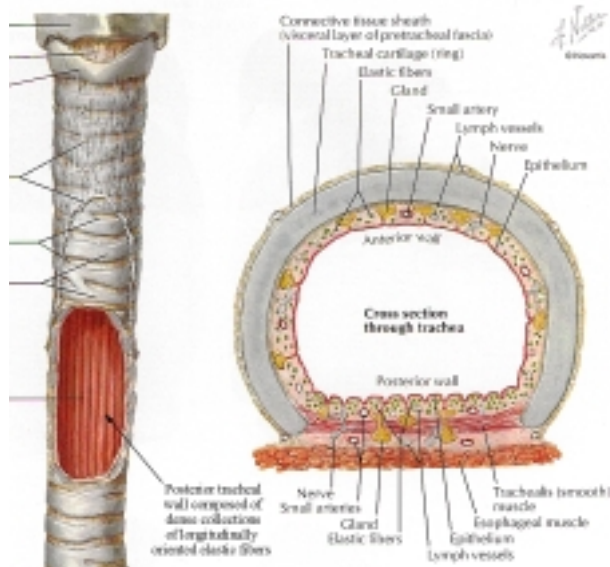


Using noncausal digital filters, we now can calculate position and velocity for objects moving with frequencies above about 75 Hz



CASIS Nov. 99 p.4

Anatomy suggested that the posterior tracheal wall would be a better reflector than the anterior wall or folds



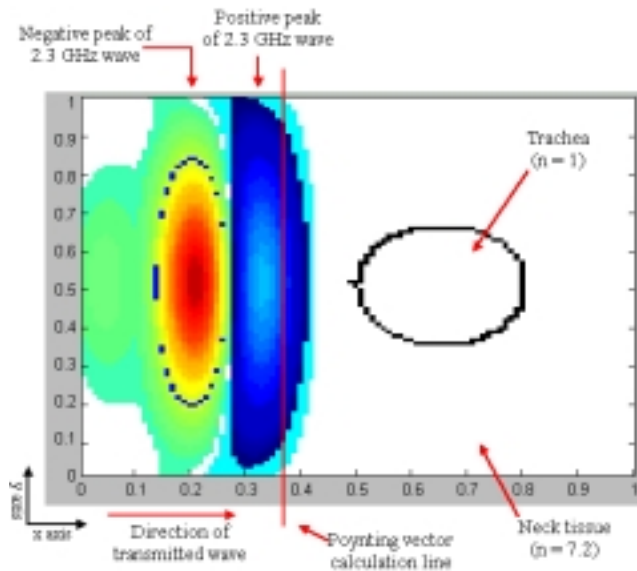
The posterior wall is flat and thick but quite flexible, while the folds do not present a significant scattering area to the GEMS

Copyright 1989. Novartis.  
Reprinted with permission  
from the *Atlas of Human Anatomy*,  
illustrated by Frank H. Netter, M.D.  
All rights reserved.

CASIS Nov. 99 p.5



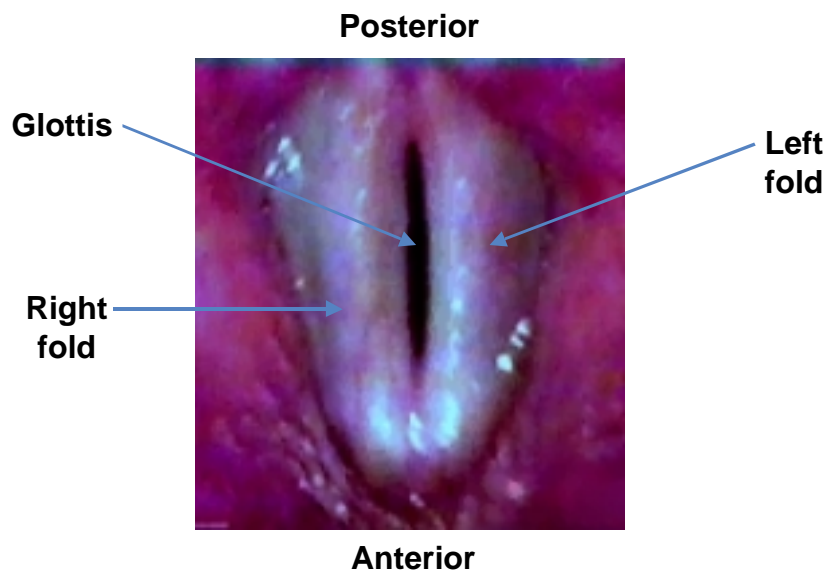
## Electromagnetic simulations using Jeff Kallman's TSARLITE indicate the trachea reflects far more energy than the max-open folds



Trachea had a reflectivity of 15.2%, more than 18x the max-open fold reflectivity of 0.8% (close to the noise level of the simulation)

CASIS Nov. 99 p.6

## High speed (1000-3000 fps) video experiments indicated fold motion did not correlate well with GEMS signal



CASIS Nov. 99 p.7



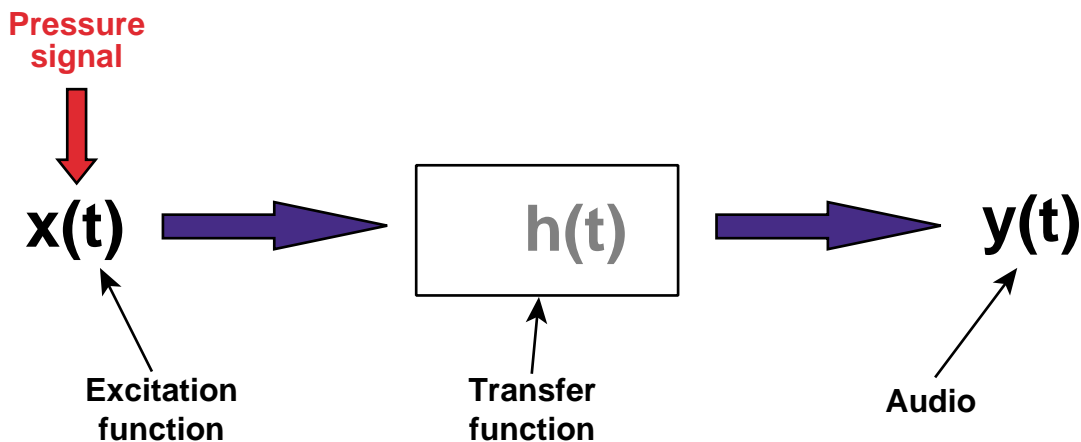
## Conclusions about physiological basis of the GEMS return



- The object responsible for the detected vibration is the subglottal posterior tracheal wall
- This wall is driven by the changes in subglottal air pressure as voicing occurs
- To define an excitation function for the human vocal tract we must relate the detected tracheal motion to the driving subglottal pressure

CASIS Nov. 99 p.8

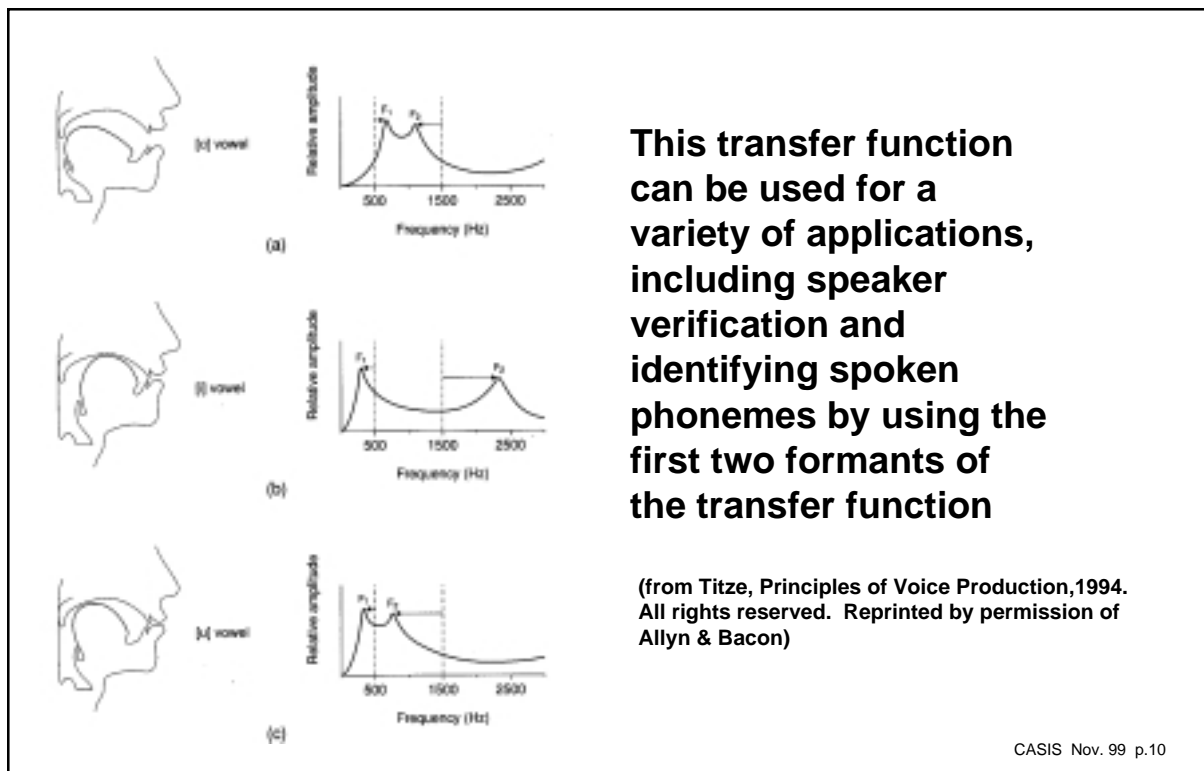
## Using the excitation function as the input to an LTI system we can calculate the transfer function of the vocal tract



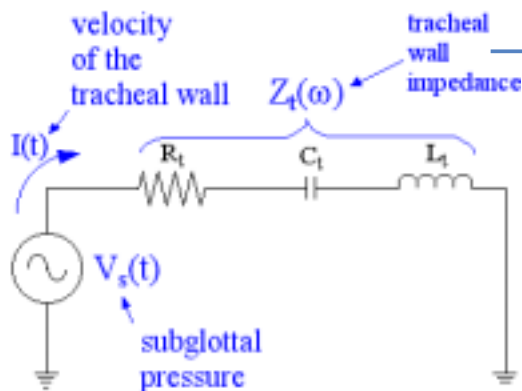
$$H(z) = Y(z)/X(z)$$

CASIS Nov. 99 p.9





A lumped-element circuit model is used to quantify the trachea's response to changing air pressure



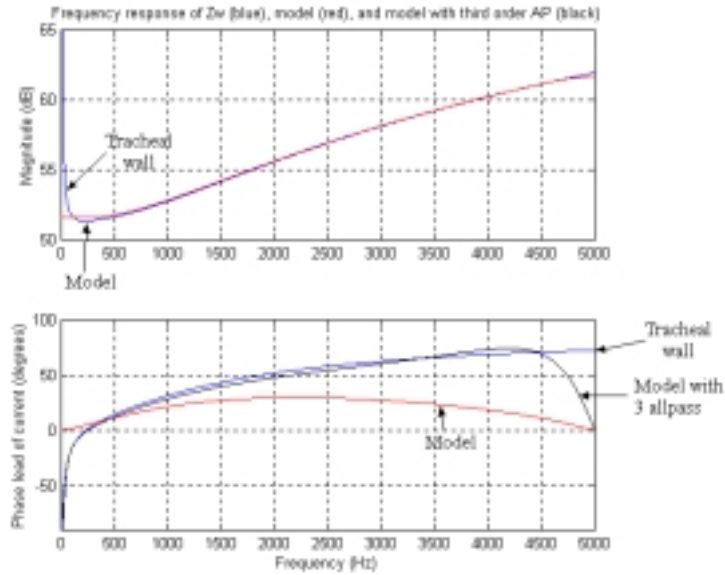
$$V_s(t) = I(t) \cdot Z_t(\omega)$$

The tracheal wall impedance was approximated by using values for several different types of human tissue. No mechanical study of the tracheal wall has been accomplished.

CASIS Nov. 99 p.11

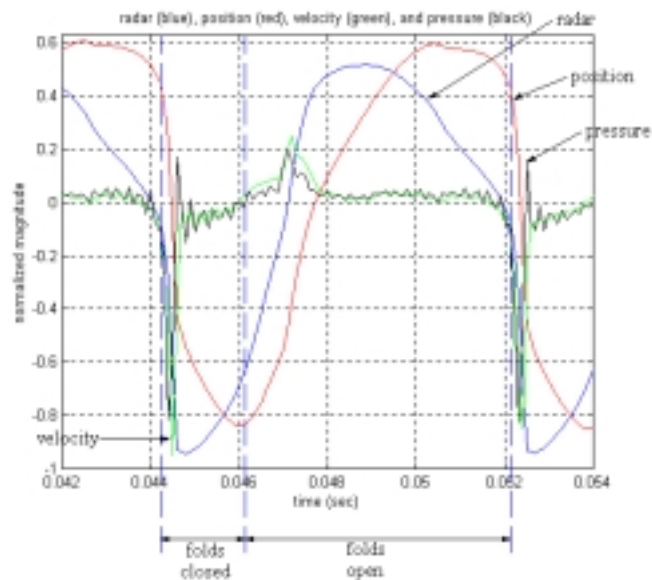


The resulting impedance frequency response is modeled with a digital lowpass and three anticausal allpass filters



CASIS Nov. 99 p.12

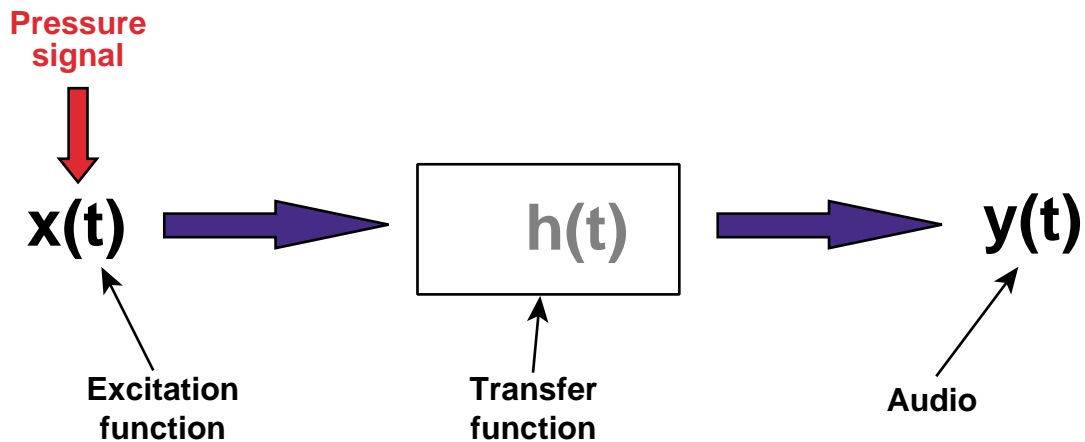
The resulting digital filter can then be used to filter the calculated wall velocity to recover the driving subglottal pressure (black trace)



CASIS Nov. 99 p.13



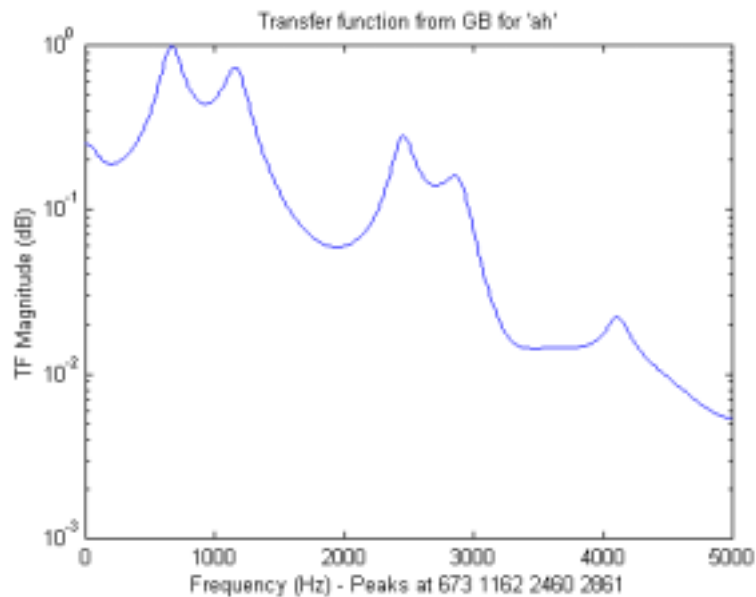
Using the excitation function as the input to an LTI system we can calculate the transfer function of the vocal tract



$$H(z) = Y(z)/X(z)$$

CASIS Nov. 99 p.9

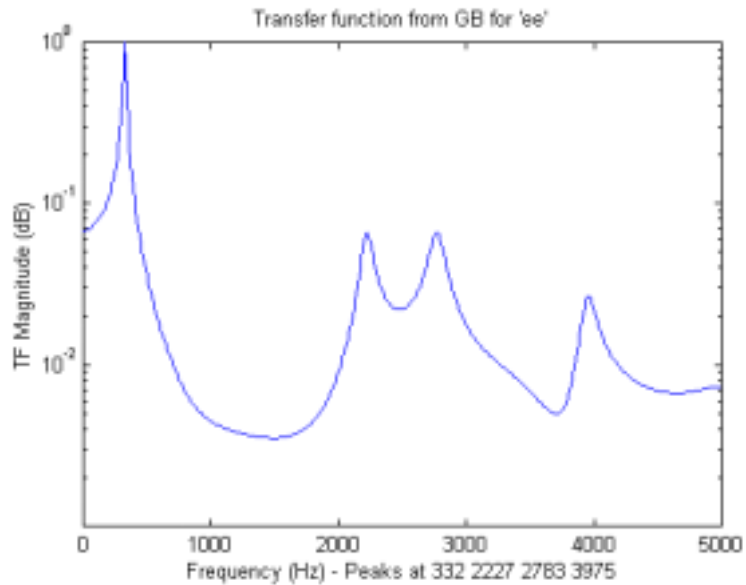
Example of the calculated TF for an “ah”



CASIS Nov. 99 p.15

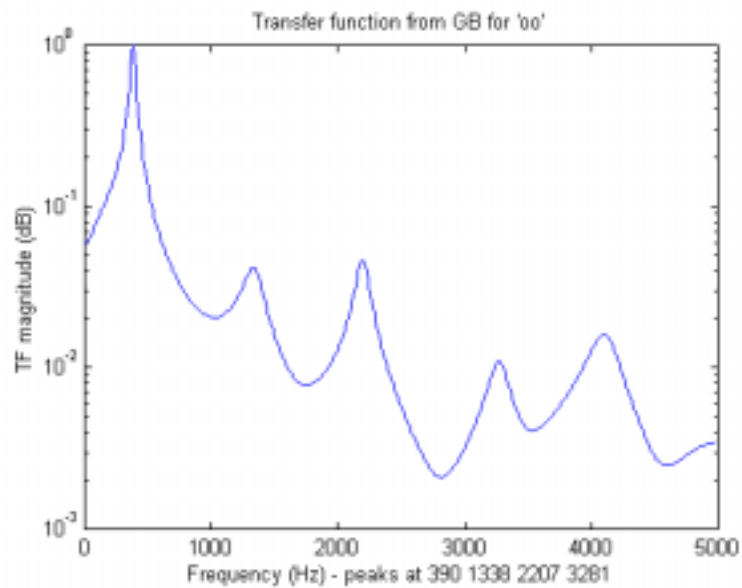


### Example of the calculated TF for an “ee”



CASIS Nov. 99 p.16

### Example of the calculated TF for an “oo”



CASIS Nov. 99 p.17



## **In conclusion**



- **The GEMS is an excellent vibration detector that can detect the motion of the rear tracheal wall during voiced speech**
- **This position of the rear wall may be differentiated and filtered using a lumped element model of the wall in order to derive the subglottal pressure**
- **This subglottal pressure may be used as the excitation function for the vocal tract**
- **With the excitation function, the transfer function of the vocal tract can be calculated**
- **The transfer function can be useful in phoneme identification and speaker verification**

CASIS Nov. 99 p.18



# Synapse TAP™ Universal Access Technology

*Marty Tibor, Synapse  
Speech Recognition and Adaptive Technology*

The Synapse TAP™ provides continuous speech recognition on computers regardless of platform. The TAP™ is a universal access technology developed in collaboration with Stanford University.

1. The user dictates text and commands at speeds up to 160 words per minute into the speech-accessing device.
2. The speech accessor converts in real-time the users' speech into keystrokes, commands and mouse events. The accessor sends its data stream to the Synapse TAP™ module.
3. The TAP™ module then converts the events in real-time into the native format of the target machine.

Presently TAP™ modules support any operating system or application found on HP, Sun and SGI UNIX workstations, Mac and IBM personal computers and any mainframe than can support a PC or Mac terminal through 3270 or other emulation. Besides providing continuous speech recognition on platforms not currently supported the TAP devices have these other features:

- Users may control with a single speech accessor up to four computers at once even if the target machines are on different hardware or software platforms.
- TAP technology collaborates numerous input technologies and therefore a user can deploy speech recognition, head tracking, single switches, foot pedals or mouse at the same time without clogging the input stream of the target computer.
- A universal command structure can easily be developed using TAP™ technology where all computers can respond to exactly the same commands. For example dictating "print document" can mean the same thing to all target computers regardless of platform or application. Training time can be greatly reduced since a user would only need to become familiar with one command structure for all environments.

Presently TAP™ technology is used a productivity tool for data entry, dictation and controlling graphical applications such as mapping, CAD and modeling. TAP™ technology is also deployed to support users with repetitive strain injuries or other disability that use advanced computer workstations.



---

# Computed Tomography

---



# Confocal Ultrasound Imaging Through Acoustically Thick Media

*UCRL-JC-136318 Abs*

*Sean K. Lehman, David D. Scott, Waleed S. Haddad*

To date, all confocal applications, whether optical or acoustic have been through optically or acoustically, respectively, thin media. We have developed a theory of transmission mode confocal imaging through thick media. We have used a full finite difference time domain simulation of acoustic field scattering through a medium to demonstrate a proof-of-principle for the theory. We present preliminary results based upon these simulations.

## Confocal Ultrasound Imaging Through Acoustically Thick Media



**Sean K. Lehman  
David D. Scott  
Waleed S. Haddad  
David H. Chambers**

**CASIS Workshop**

**November 11, 1999**



## **Introduction**

---



- **Definition of Problem**
- **Problem Domain**
- **Scanning Technique**
- **Simulated Examples**
- **Current Results**

## **Definition of Problem**

---

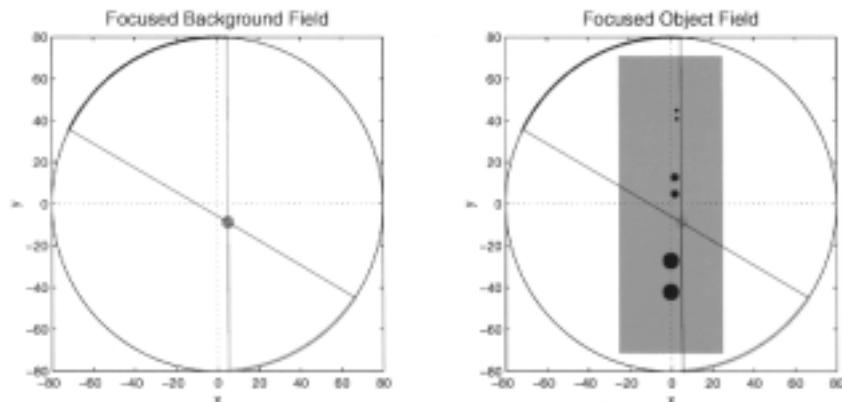


- **Perform an acoustic confocal scan of an object and construct an image of the object using tomographic techniques;**
- **Assume low sound speed contrast, i.e. small changes in refractive index.**

**We hope to achieve higher contrast reconstructions over traditional straight ray tomographic techniques.**



## Problem Domain



## Scanning Technique



- Perform a scan through the object at the desired resolution. This is the inhomogeneous received field.
- Perform a background scan (in the absence of the object) at the same resolution. This is the homogeneous received field.
- At each focus, correlate the inhomogeneous received field with the homogeneous received field.
- Differences between the homogeneous and inhomogeneous received field amplitude and arrival times are related to the object's attenuation and sound speed at the focus..

We are developing a tomographic reconstruction algorithm which uses these measured differences to image the object's absorption and refractive index distribution.

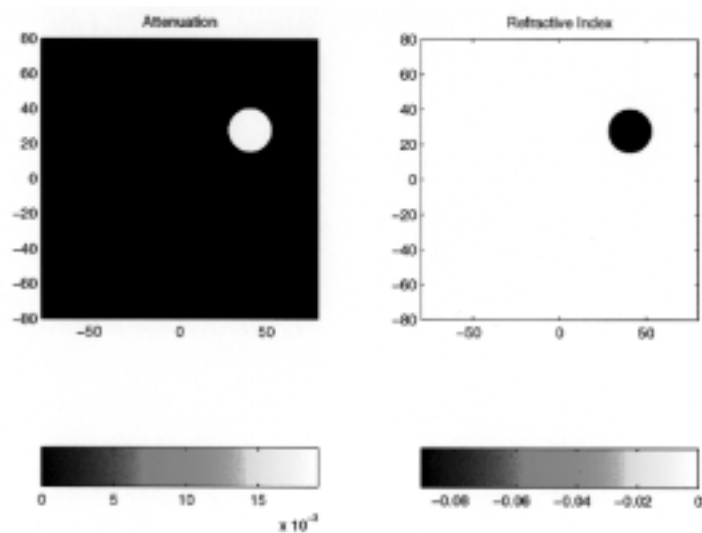


## Simulated Examples



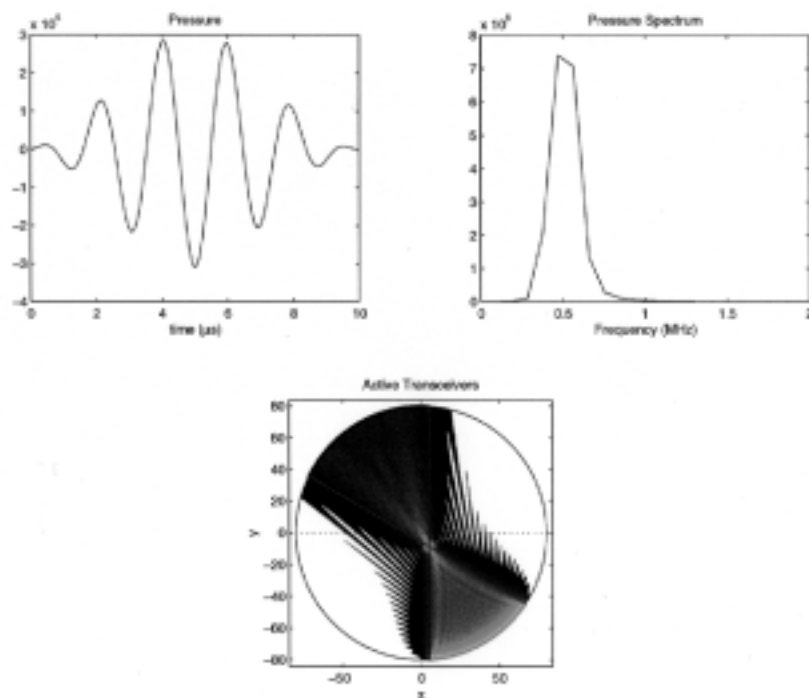
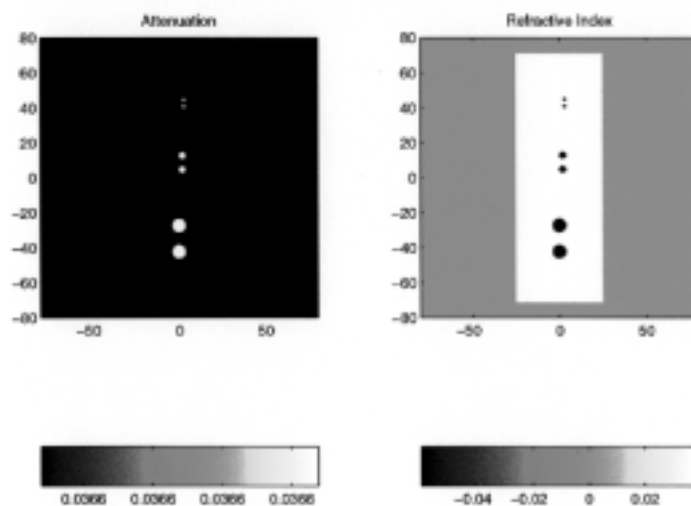
- Finite difference time domain simulation of object in a ring scanner with 360 equally spaced transmitters interleaved with 360 equally spaced receivers.
- 0.5 MHz

## Simulated Objects

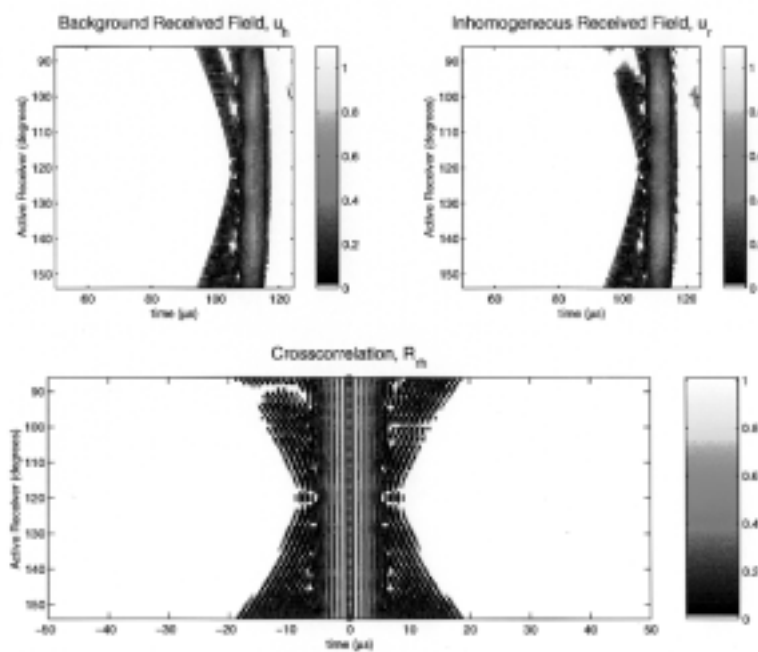
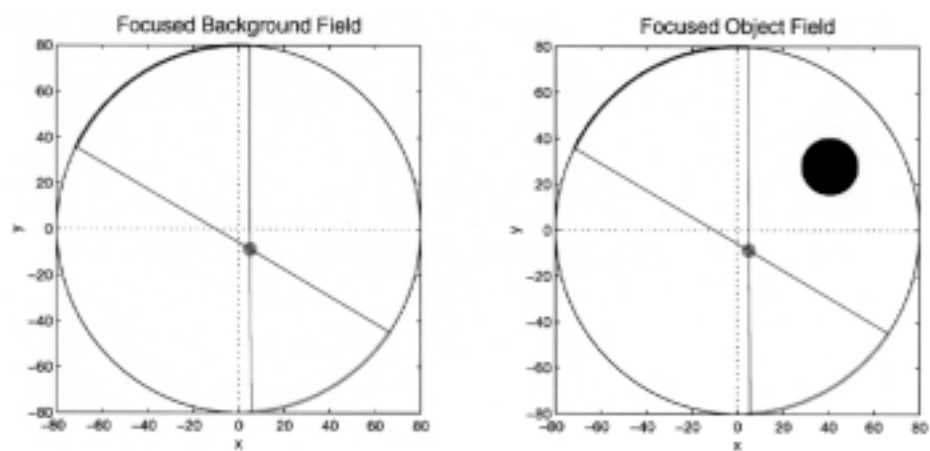




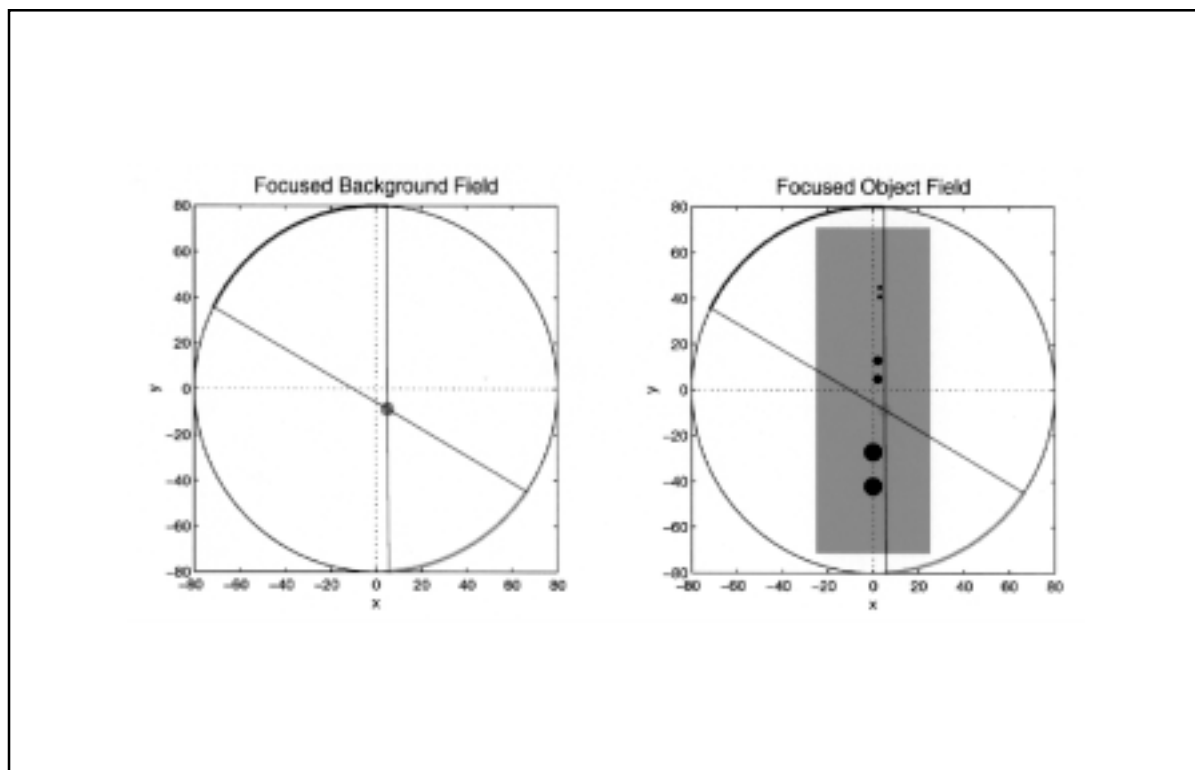
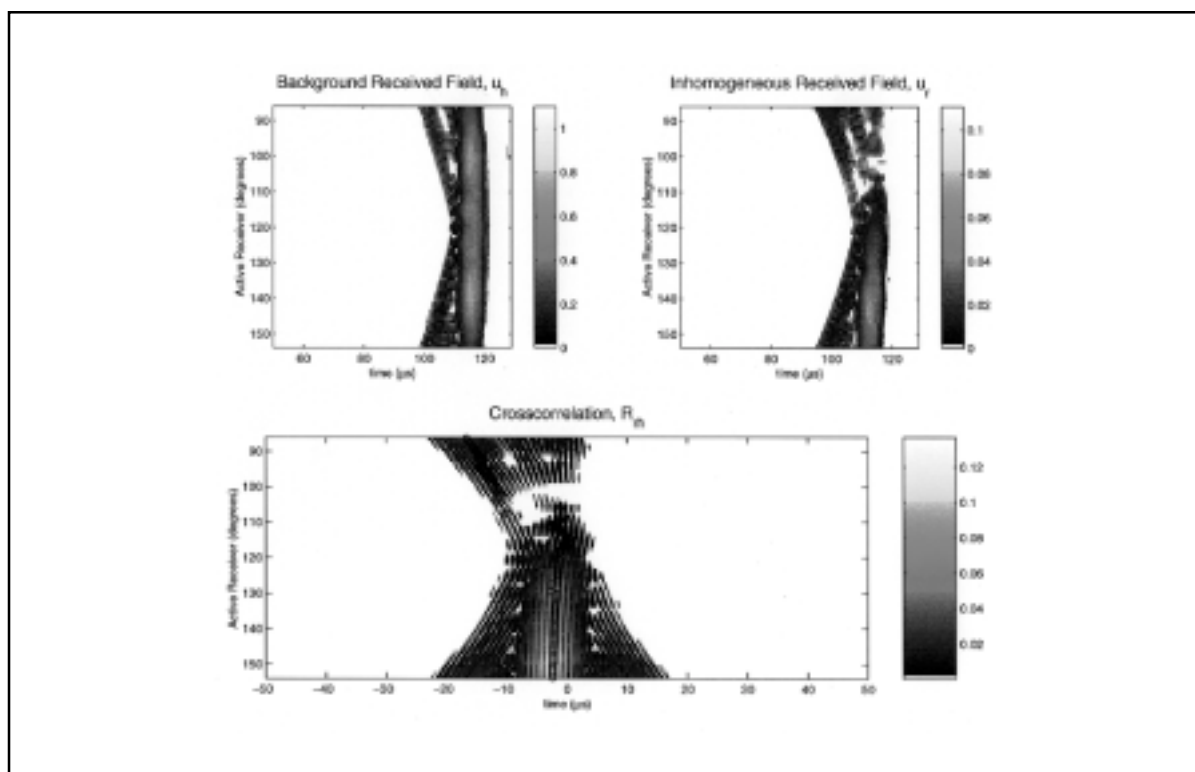
## Simulated Objects



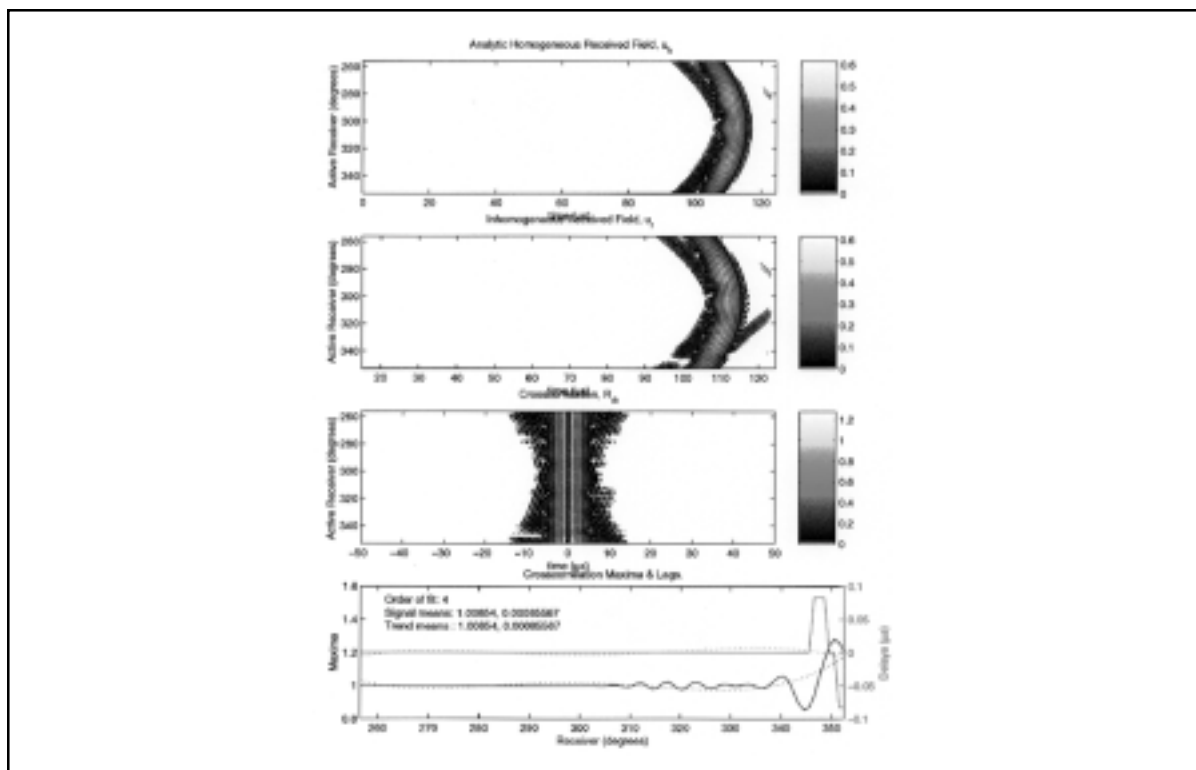
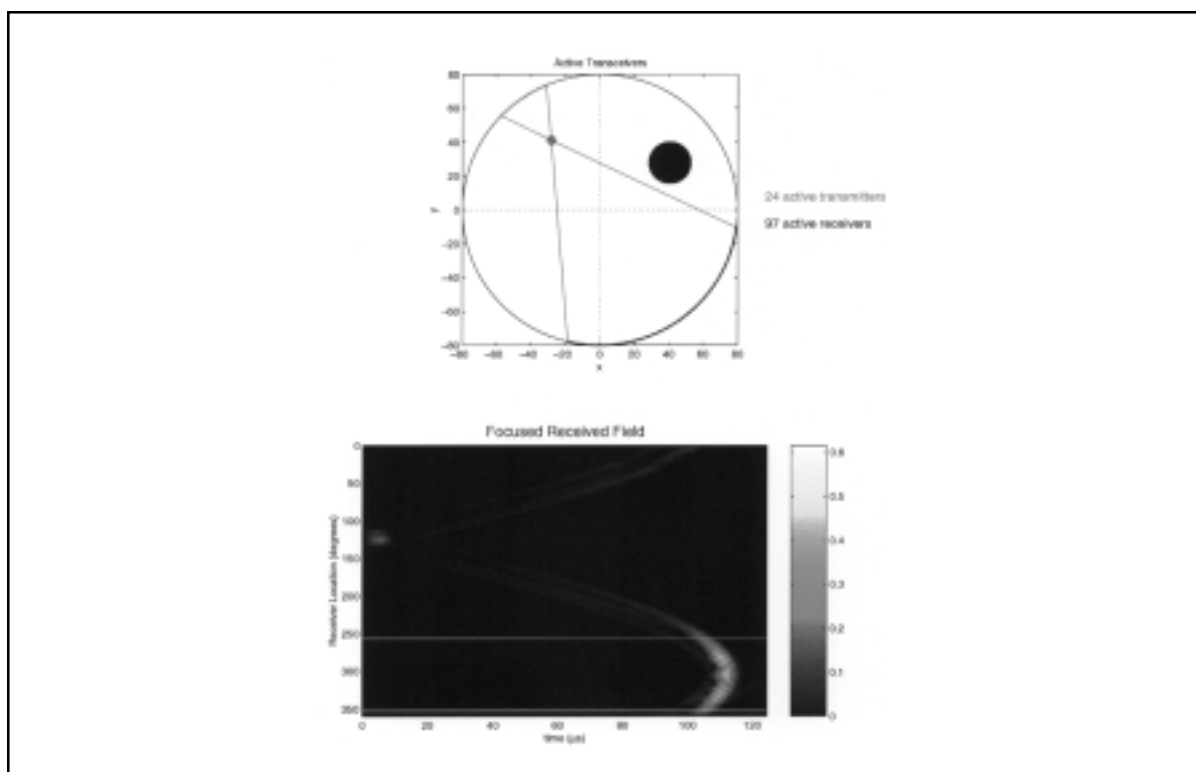




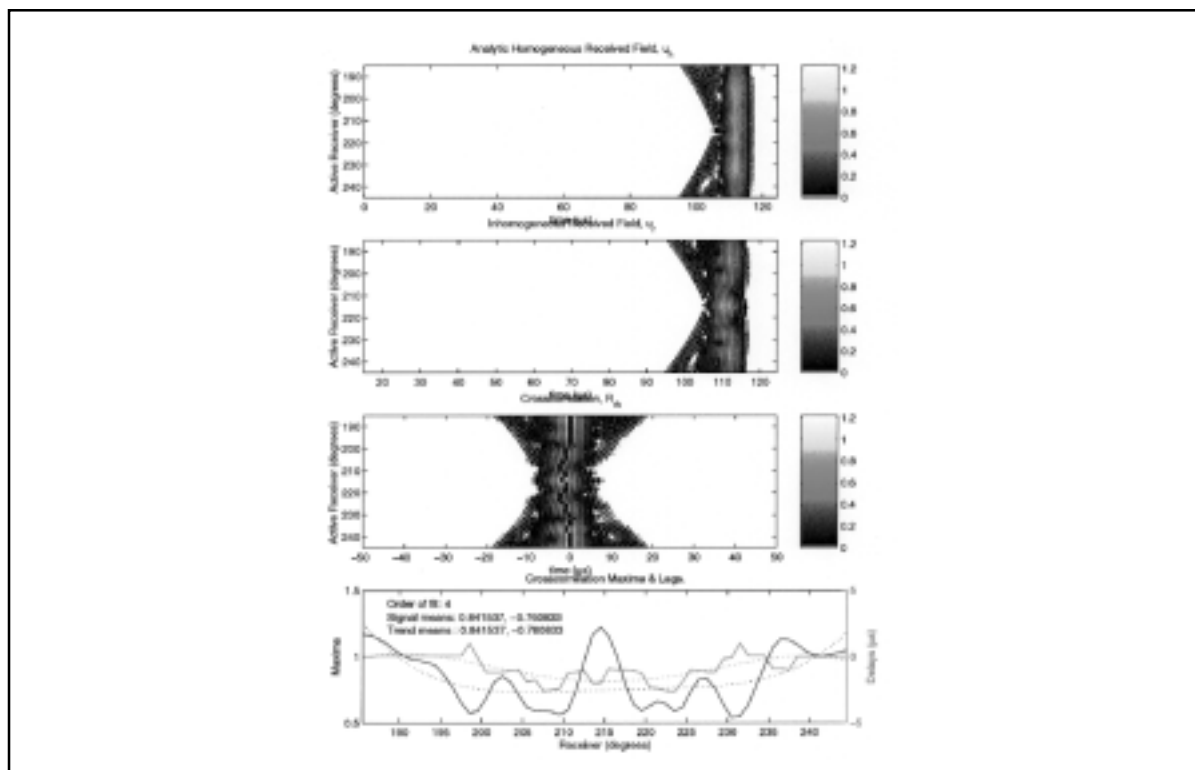
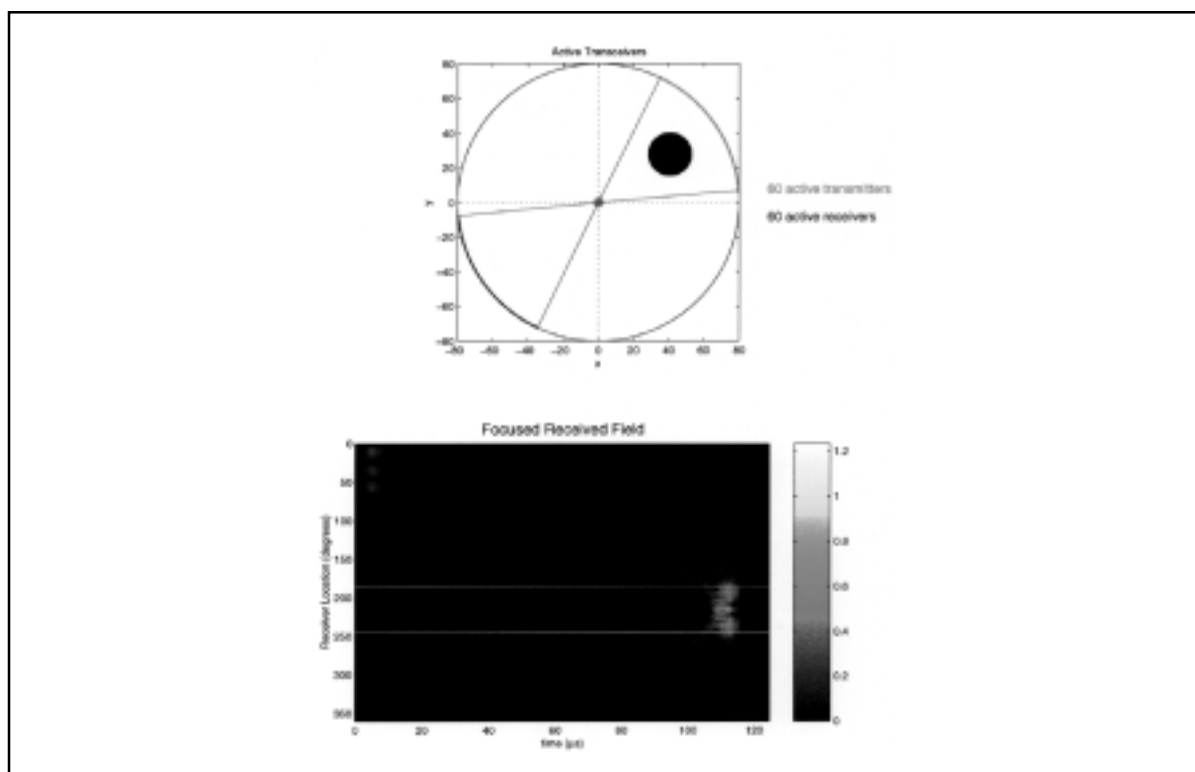




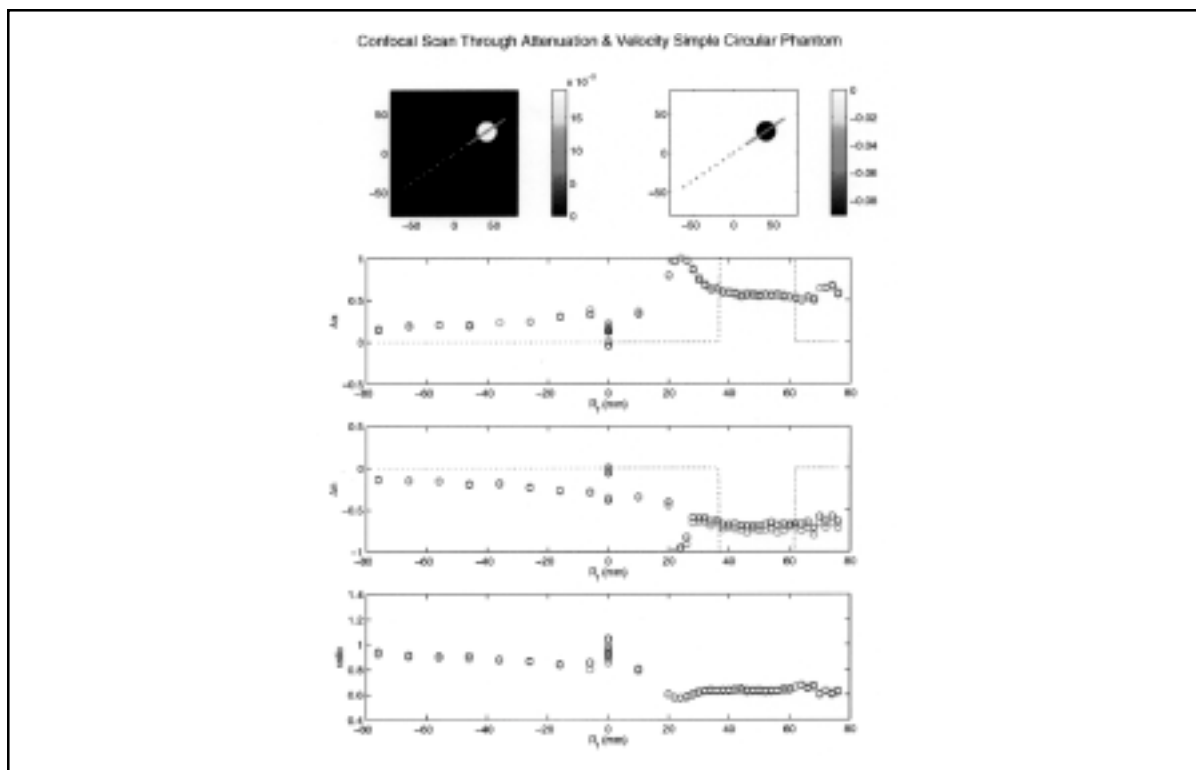
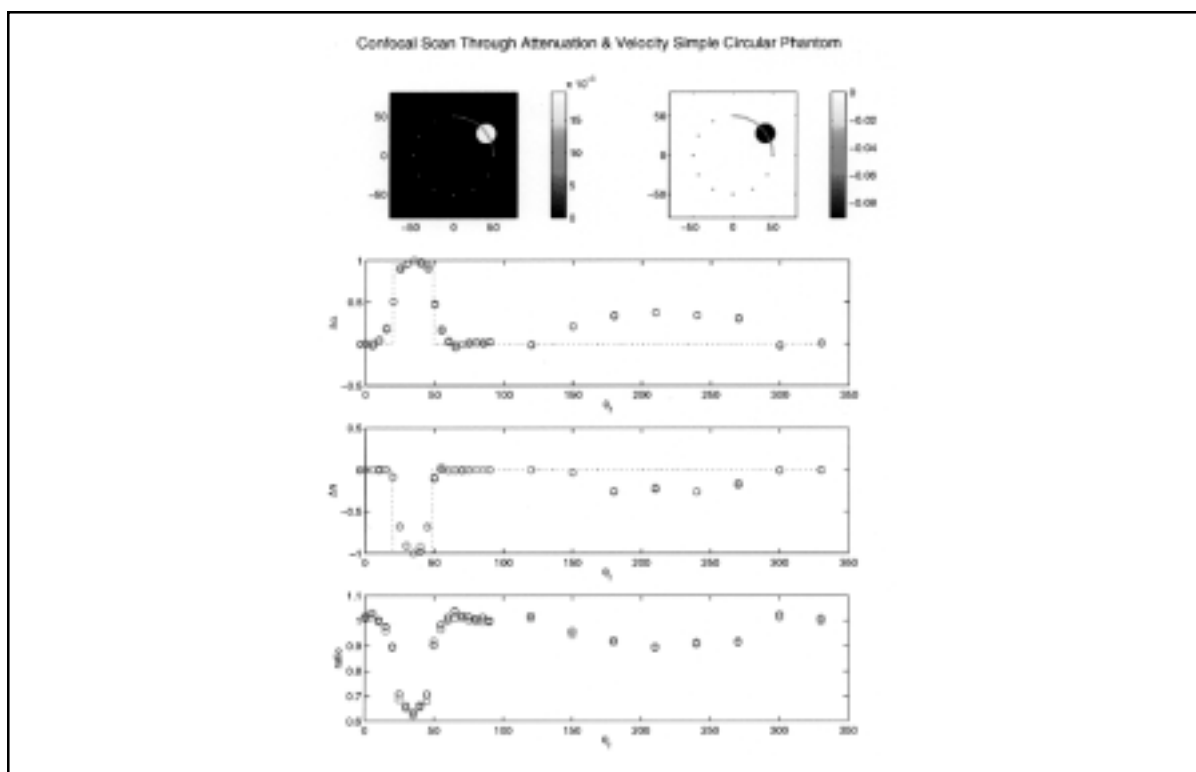




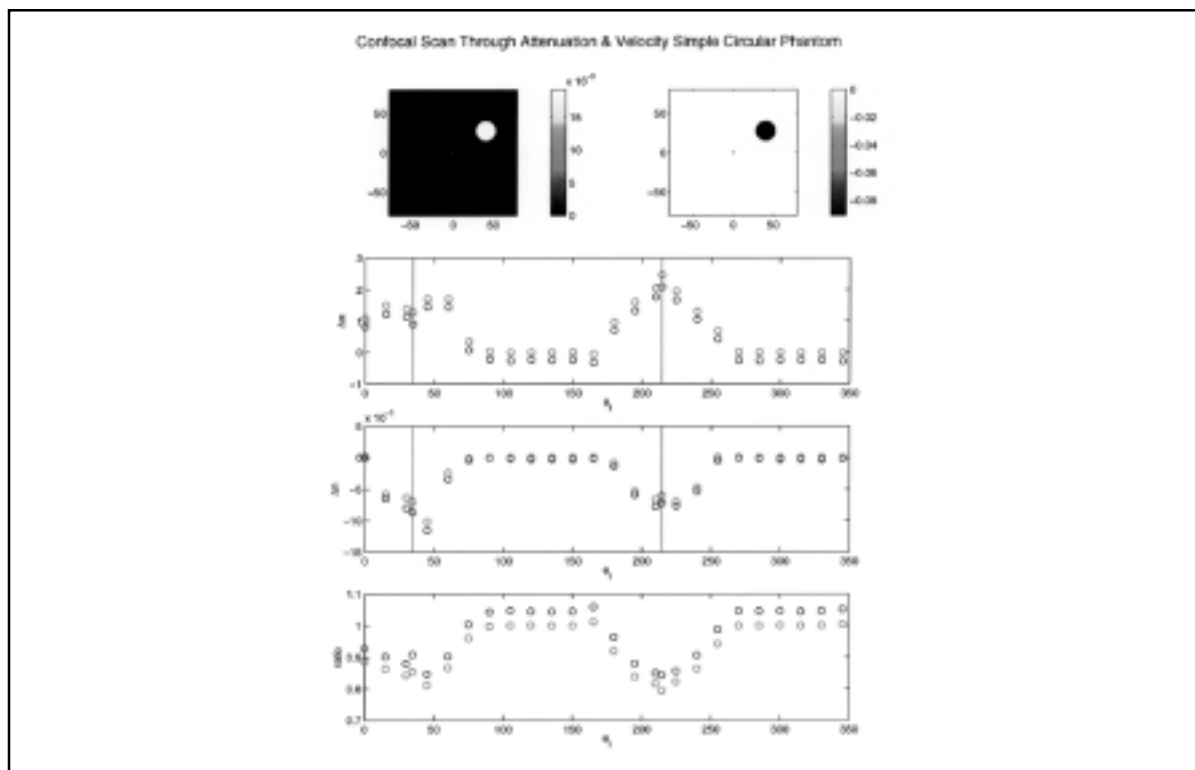
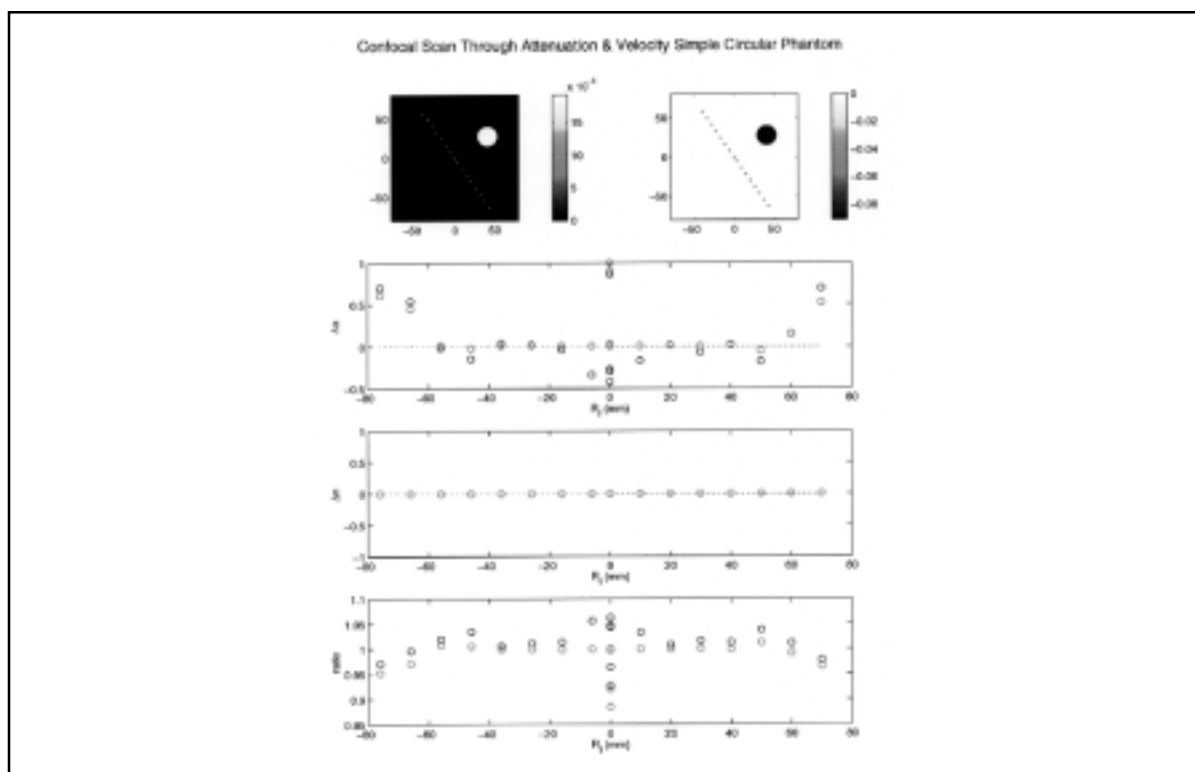












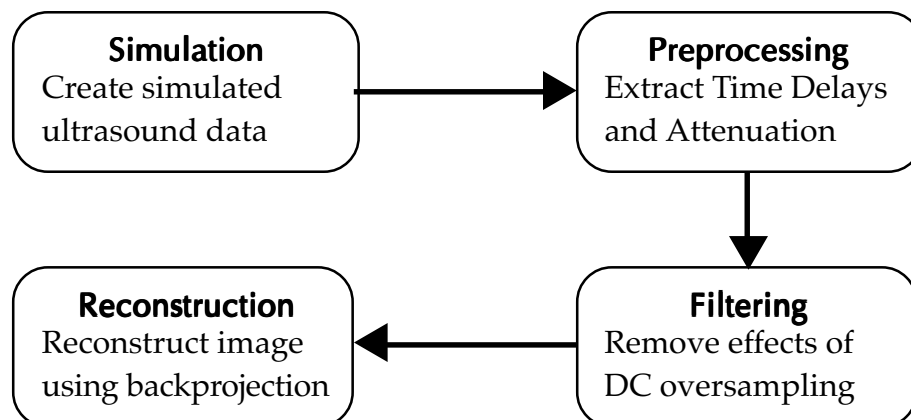


# Filtered Backprojection for Low Contrast Medical Ultrasound Tomography Using Matlab

CASIS Workshop  
November 11–12, 1999

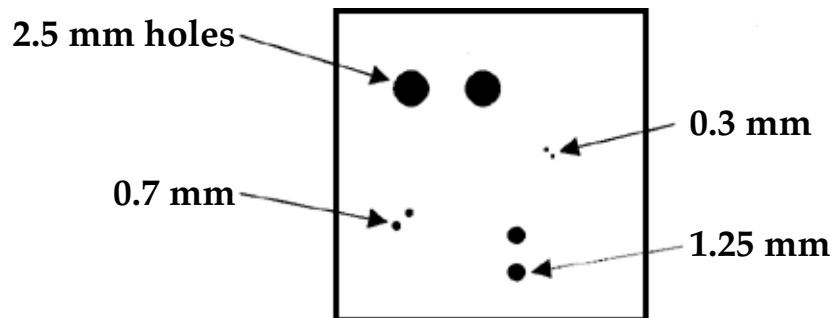
David Scott  
Sean Lehman  
Lee Haddad  
David Chambers

## Image Reconstruction Protocol





## A Simple Test Object

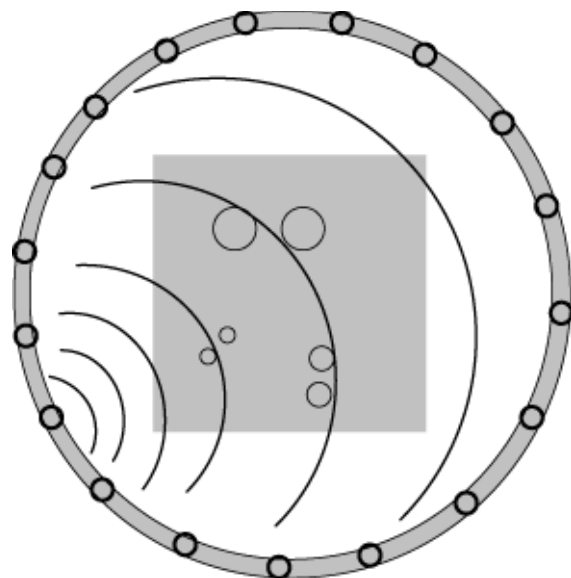


## Simulating Ultrasound Data

Simulates 2D and 3D  
ultrasound propagation

Model is accurate for  
reconstructing transmission  
and reflection modes images

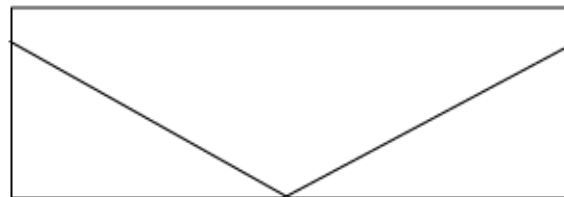
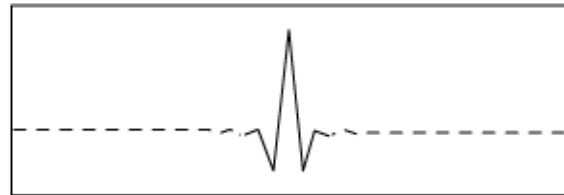
Geometry is easily  
configurable



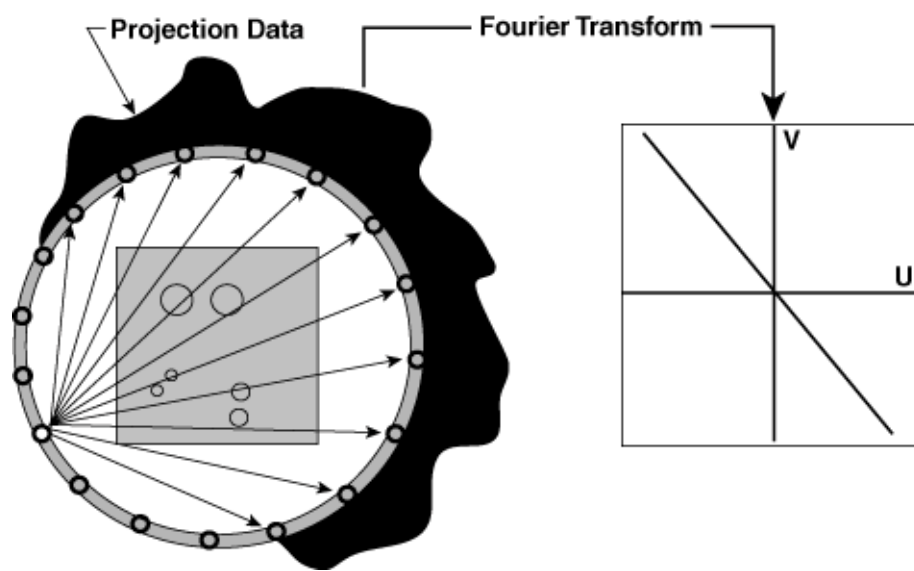


## Filtering the Projection Data

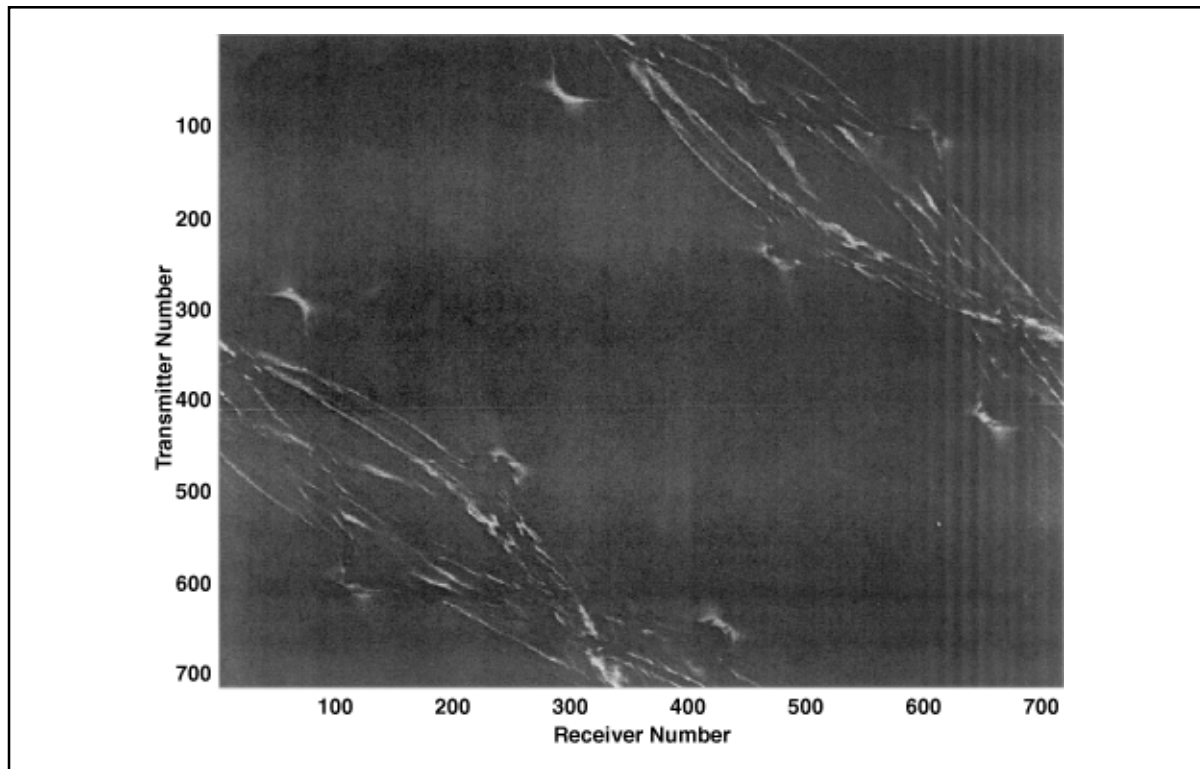
The “ramp” filter reduces the effect of oversampling the lower frequencies.



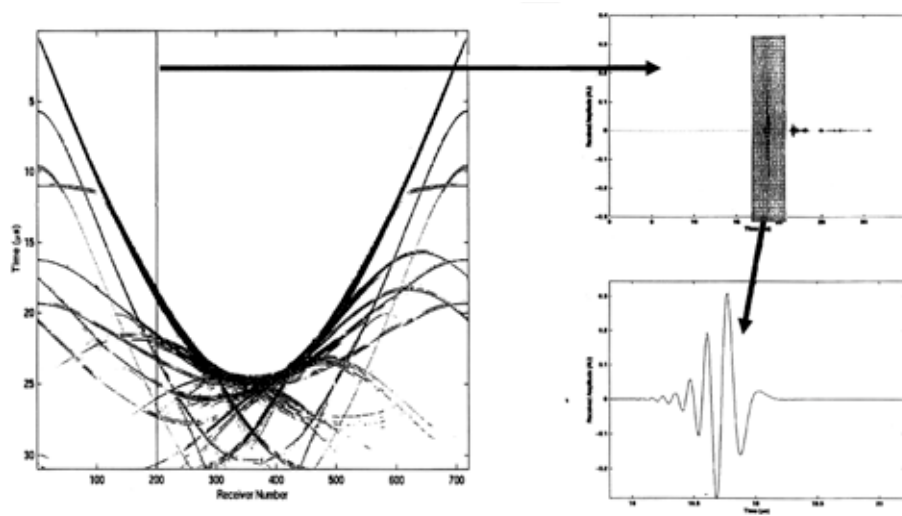
## Filtered Backprojection: The Fourier Slice Theorem



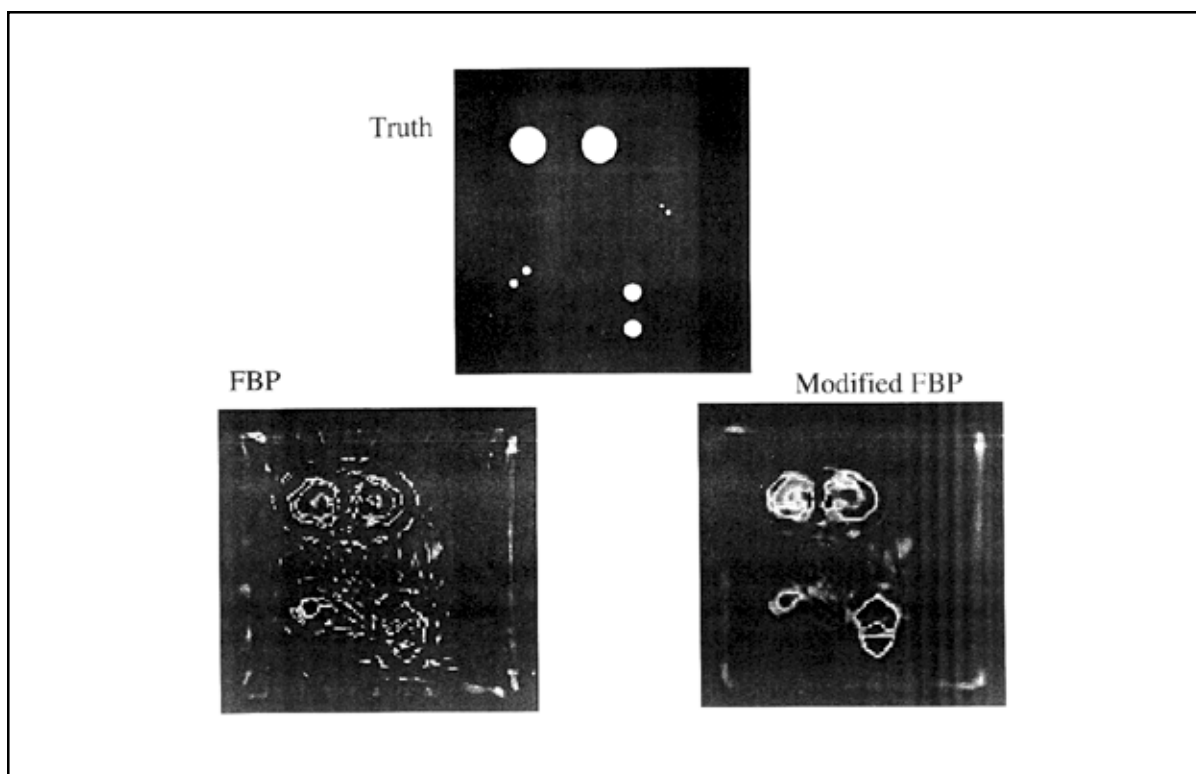




## Extracting the Time Delay and Attenuation









## Optimal Beam Hardening Correction for X-Ray Tomography

*Jesse Kolman*

Images produced by x-ray tomography using a high-energy broadband source are distorted by beam hardening, which is the result of the fact that the attenuation of x-rays varies with wavelength.

If the object being imaged consists of a single material, the effects of beam hardening can be removed effectively.

However, for objects composed of multiple materials, the image reconstruction problem is inherently ill-posed and cannot be solved without a priori information.

We have developed an algorithm which is capable of reconstructing an image of an object containing multiple materials when a priori knowledge is available. The algorithm uses a constrained optimization algorithm to find the maximum likelihood estimate of the object.

In general, the algorithm can be expected to be successful when the a priori information reduces the ill-posed nature of the problem adequately. In the absence of such information, the algorithm returns a solution which may be suboptimal, since the inherent ambiguities cannot be resolved. In such cases, useful images can often still be obtained, at a high computational cost.

*keywords: beam hardening, x-ray tomography, optimization Algorithms*

## Optimal Beam Hardening Correction for X-Ray Tomography



**Jesse Kolman**

**November 11, 1999**



## **High Energy X-Ray Tomography**



- **High energy (short wavelength) x-rays are needed to penetrate dense materials**
- **Sources of high energy x-rays are typically broadband**
- **These conditions result in beam hardening**

## **Beam Hardening**



- **X-rays with high energy are attenuated less than those with lower energy**
- **Transmitted x-rays have a higher relative concentration at high energies**
- **Attenuation is material dependent**



## Beam Hardening Model



- The measurements are related to the object by

$$\hat{y}_i = \sum_k s_k \exp \left\{ -\sum_m \alpha_{mk} A_i \underline{x}_m \right\}$$

where

$\underline{x}_m$  = Sources of high energy x-rays are typically broadband

$A_i \underline{x}_m$  = These conditions result in beam hardening

$s_k$  = Sources of high energy x-rays are typically broadband

$\alpha_{mk}$  = These conditions result in beam hardening

## Single Material Algorithm



- If the object consists of a single material, the model becomes

$$\hat{y}_i = h(A_i \underline{x})$$

where

$$h(x) = \sum_k S_k \exp\{-\alpha_k x\}$$

- Inverting this pointwise nonlinearity results in a standard linear inverse problem



## Optimization Problem



- If the measurement noise is independent identically distributed Gaussian noise, then the maximum likelihood estimate of the object is the minimizer of

$$f(\underline{x}) = \frac{1}{2} \sum_i (\hat{y}_i - y_i)^2$$

Subject to any known constraints

## Optimization Algorithm



- We use a constrained conjugate gradients algorithm, which iteratively searched for a solution using the gradient

$$\nabla_{\underline{x}_l} f = A^T u_l$$

where

$$u_{li} = (\hat{y}_i - y_i) \sum_k S_k \alpha_{mk} \exp \left\{ - \sum_m \alpha_{mk} A_i \underline{x}_m \right\}$$

- The algorithm dynamically enforces various constraints including nonnegativity



## **Difficulties**

---



- Typically, the number of object pixels is approximately the same as the number of measurements
- With one variable per material per pixel, the problem is ill-posed for multiple materials
- Many objects can result in the same set of measurements

## **A Priori Knowledge**

---



- To make the problem tractable, inherent ambiguities must be reduced using a priori information
- Prior knowledge must be accurate and possible to enforce as constraints on the variables



## **Region of Support Constraints**

---



- **Certain materials may be known to exist only in certain parts of the image**
- **Disallowing solutions which have materials outside their regions of support effectively reduces the number of variables**
- **Good results can be obtained even with overlapping regions, as long as the number of variable is no more than the number of measurements**

## **Initial Estimate**

---



- **We can start the algorithm with an initial guess at the object based on a blueprint**
- **Deviations or defects in the object can be accurately imaged if the true object is close to the initial estimate, even with multiple materials**

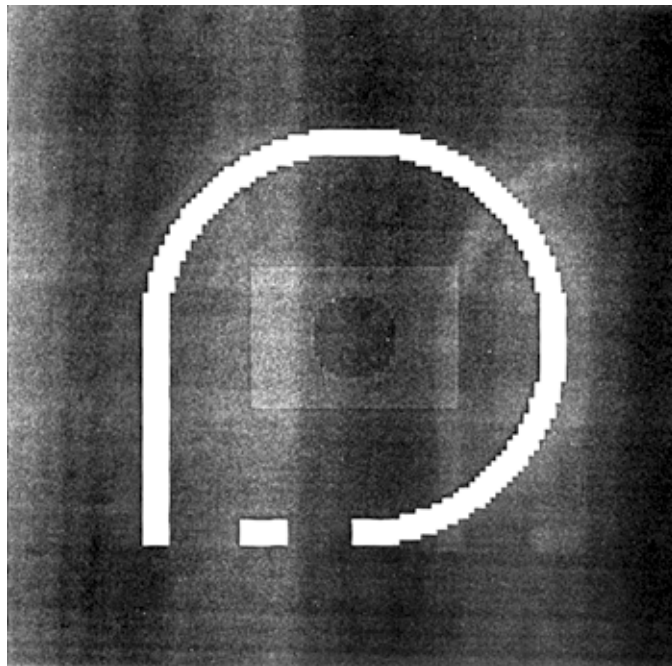


## Conclusions

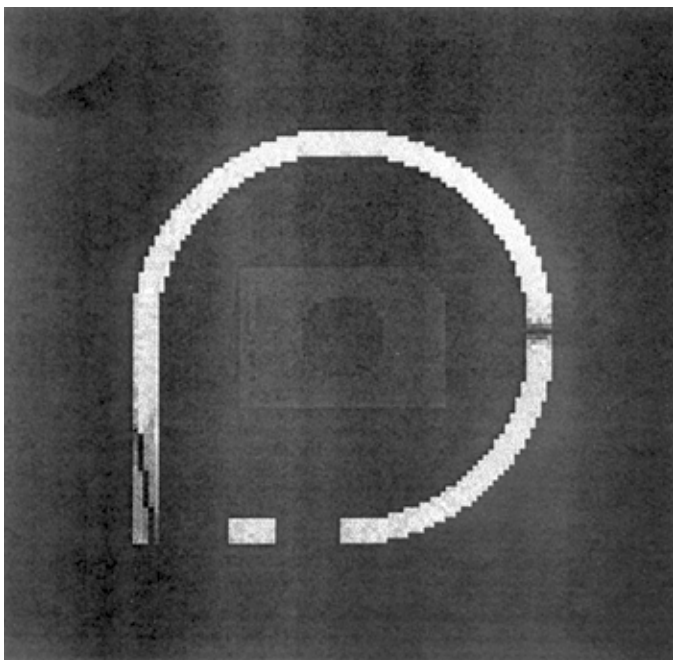
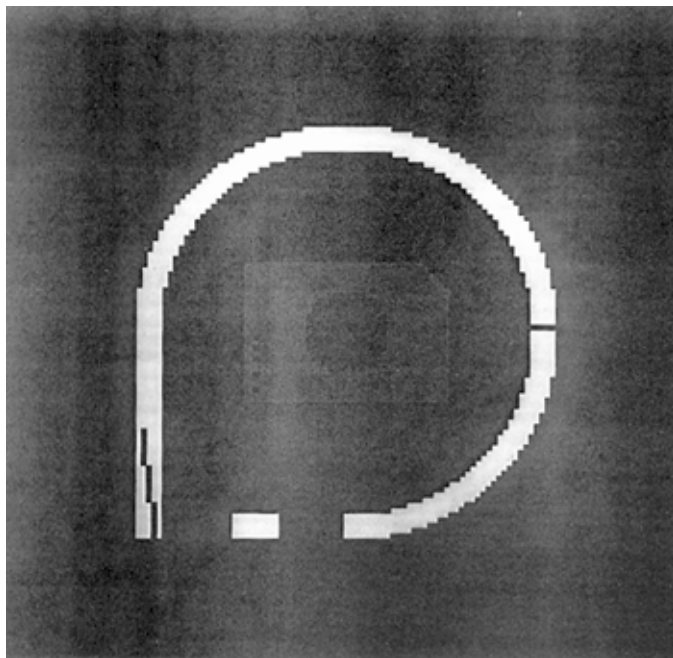
---



- If model parameters are accurately known and a priori information can sufficiently reduce ill-posedness, the problem can be accurately solved with an optimization algorithm
- Computational cost is high by manageable









# Performance Evaluation of Amorphous-Silicon Flat Panel Array for X-Ray Digital Radiography and Computed Tomography

(UCRL-VG-136259)

*K. W. Dolan, J. J. Haskins, C. M. Logan , R. D. Rikard, and D. J. Schneberk*

## SUMMARY

Lawrence Livermore National Laboratory (LLNL) has evaluated amorphous-silicon flat panel arrays for use as imaging elements in x-ray digital radiography and computed tomography systems. Digital radiography systems are being developed for applications where digital image format and fast image acquisition provide an advantage over film, when image quality and dynamic range requirements can be satisfied. Computed tomography systems based on the flat panel technologies are being evaluated as a means of improving the speed of image acquisition as well as image quality, particularly for large fields-of-view (greater than 5 or 6 inches on a side).

The flat panel amorphous-silicon (a-Si) array used in these studies provided a sensitive area of 195 mm x 244 mm (7.7 in x 9.6 in) with 2.95 million pixels, where pixel dimensions are 127  $\mu\text{m}$  x 127  $\mu\text{m}$  (0.005 in x 0.005 in). The array incorporates charge sensitive amplifiers and 12-bit A/D converters providing a dynamic range, as determined here, of approximately 3400:1. The a-Si array converts light images into electrical charge, pixel by pixel. The light images are produced by x-ray interactions in a phosphor screen placed in intimate contact with the a-Si array. Evaluations were made of the resolution and efficiency of different phosphor screens, filter and foil set combinations. Spatial resolution and contrast sensitivity were determined as a function of x-ray source energy from 70 kV to 9 MV. Dynamic range, image acquisition speed, image reproducibility, potential for latent image formation, dark current, and corrective measures for source instability were also studied. Data acquisition and image normalization software were developed for both digital radiography and computed tomography. Application examples are discussed.



UCRL-VG-136259

## Performance Applications of Amorphous-Silicon Flat Panel Imager

prepared by  
K. W. Dolan, J. J. Haskins, C. M. Logan, R. D. Rikard and D. J. Schneberk  
presented by Ken Dolan

Nondestructive and Materials Evaluation Section  
Lawrence Livermore National Laboratory



presented to CASIS Workshop  
Livermore, CA  
November 11-12, 1999

Work performed under the auspices of the U. S. Department of Energy by Lawrence  
Livermore National Laboratory under contract W-7405-Eng-48

KD\_CASIS #1

The a-Si flat panel array provides a new capability  
for radioscopy and computed tomography



Flat panel a-Si array configured for  
radioscopy and computed tomography  
with rotation/translation staging

### Features

- Image sensor type  
a-Si with conversion screen
- Image area  
195x244 mm (8x10 in)  
282x406 mm (11x16 in)
- Pixel format  
1536x1920 (2.9 M pixels)  
2232x3200 (7.1 M pixels)
- Pixel size: 127  $\mu\text{m}$
- Fill factor: 57%
- Dynamic range: >2000:1
- MTF: > 35% at 2 lp/mm
- Image transfer time:  
< 2 sec (2.9 M pixels)  
< 5 sec (7.1 M pixels)

KD\_CASIS #2



## a-Si flat panel satisfies radioscopy and computed tomography needs for large area imaging



### Advantages

- Flat profile for film replacement (1.8 in depth)
- Provides large area imaging (11x16 in) x 4 panels
- Performs well at all energies tested (120 kV to 9 MV)
- Electronics can be shielded outside image area
- Can choose phosphor screen to match x-ray flux

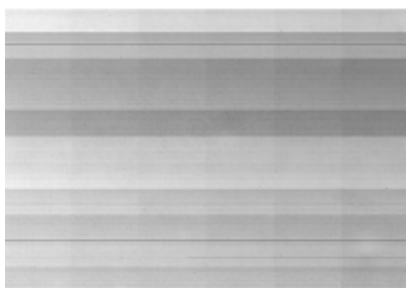
### Limitations

- Dynamic Range < 12 bits
- Fixed pixel size (127  $\mu\text{m}$ )
- Readout electronics susceptible to radiation damage
- Each bank in array has different gain (128 rows/bank)
- Each pixel needs to be calibrated for each use
- Frame transfer time (2 - 5 sec)\*

\* Smaller array (179x238 mm) provides real-time (25-30 frames/sec) at 2x2 binned

KD\_CASIS #3

## Dark field and light field image corrections are needed to obtain calibrated image



Dark field image at no x-ray fluence

### Dark Field Image

Pixels start at maximum value (3700 - 4000)  
128 pixel row elements per bank  
Gain or offset differences between banks  
Small additive bias across panel  
A few bad pixels are also part of the image



Light field image at maximum x-ray fluence

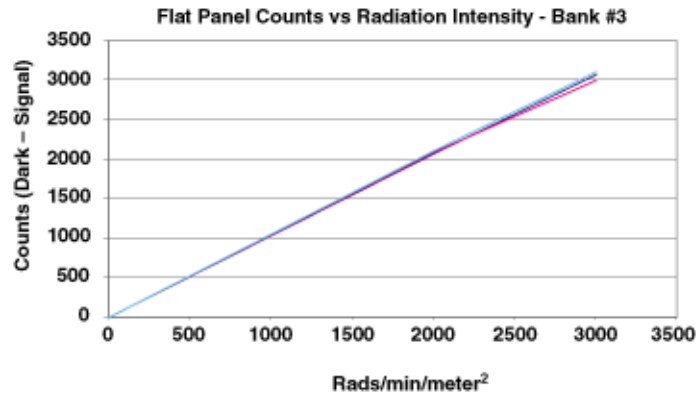
### Light Field Image

Panel responds to light by counting down  
Maximum usable pixel value is approx. 500  
Dynamic range is variable between banks  
Clipping in corners of this image is caused by collimation

KD\_CASIS #4



### Gain response is bimodal linear and varies slightly by pixel and bank

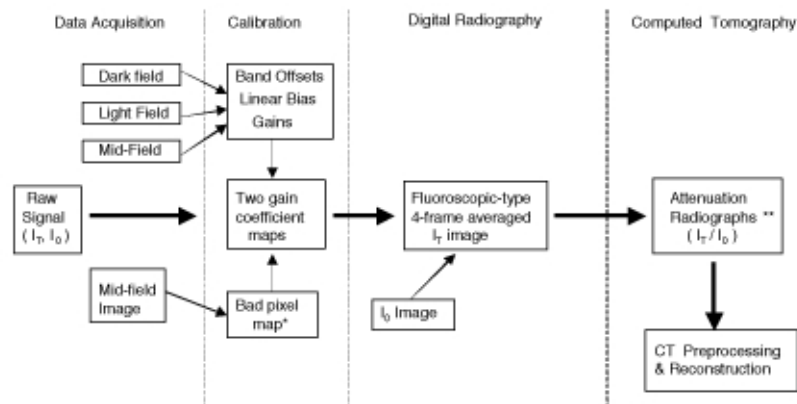


Plot shows pixel digital readout for 3 pixels in bank # 3 as a function of x-ray exposure.

- Small but significant gain difference between pixels.
- Bimodal linear gain with change in slope at approx. 2000 - 2200 counts

KD\_CASIS #5

### LLNL processing protocol for digital radiography and computed tomography



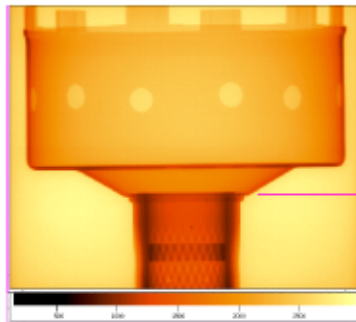
\* Bad pixels are corrected by 6-point average of surrounding good pixels

\*\* Attenuation radiographs are unclipped & unthresholded

KD\_CASIS #6

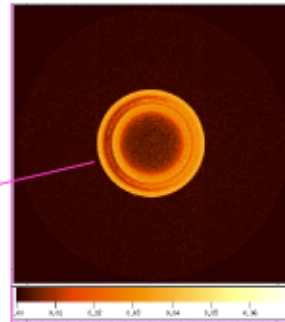


LLNL data analysis protocol has produced valid  
computed tomography results



Digital radiograph of prototype  
clutch retainer

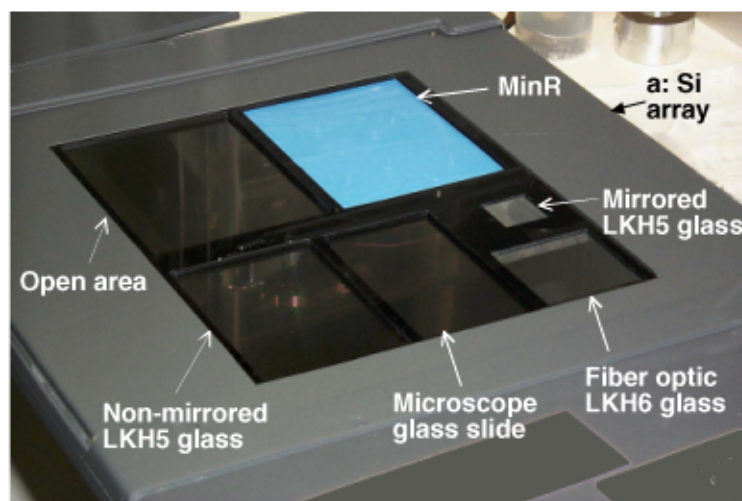
Slice plane



Computed tomography slice  
at laser weld joint

KD\_CASIS #7

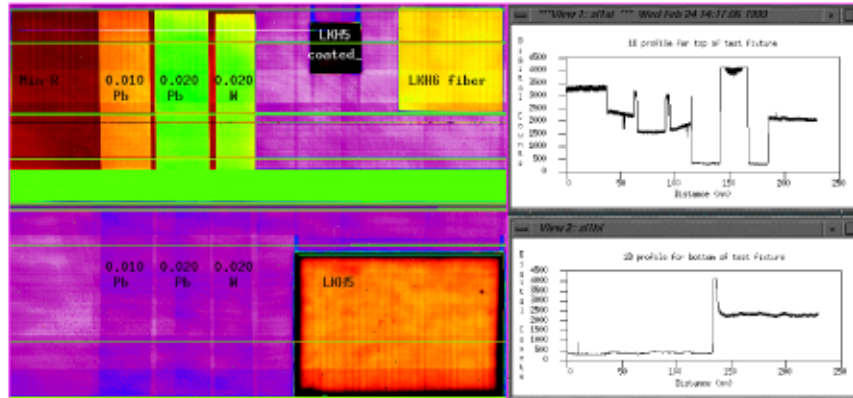
An array of phosphor screens and scintillators  
were tested for signal conversion and resolution



Mosaic arrangement of screens and scintillators in a: Si array imaging fixture



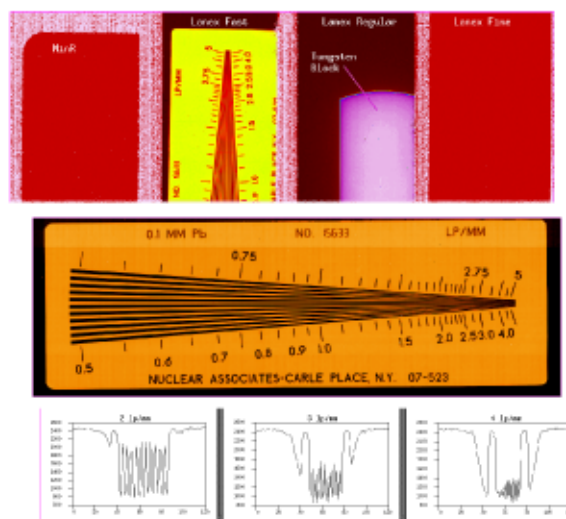
## Phosphor screens with a: Si outperformed scintillators and metal foil/screen combinations



Best performance was obtained with Kodak MinR (Lanex Fine) phosphor screen

KD\_CASIS #9

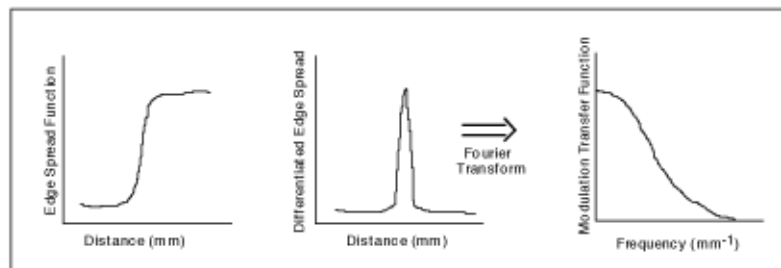
## Resolution and MTF were determined from digital images of line pair gauges and block edges



KD\_CASIS #10



Modulation transfer function (MTF) was determined by the edge spread method

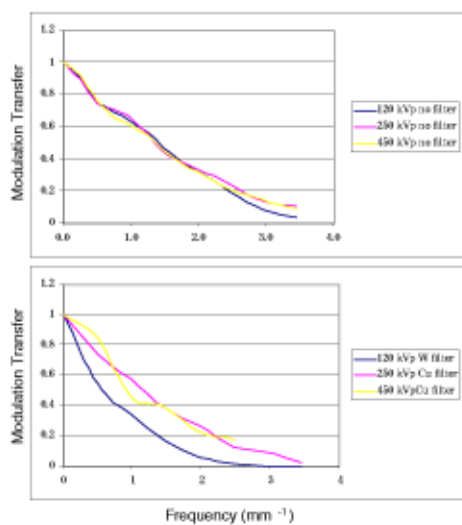


Schematic of edge spread method for determining MTF

MTF provides contrast as a function of feature size

KD\_CASIS #11

MTF was determined with MinR screen for filtered and unfiltered x-rays at 120, 250 and 450 kVp



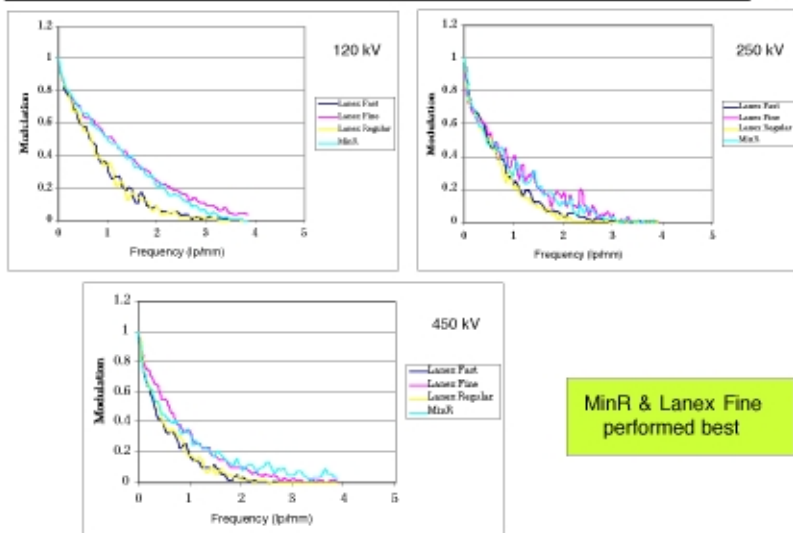
MTF is greater than 5% at 3.5 lp/mm  
For unfiltered spectra

MTF is greater than 5% at 3.0 lp/mm  
120 kV and 250 kV, greater than 5%  
at 2.0 lp/mm for filtered spectra

KD\_CASIS #12

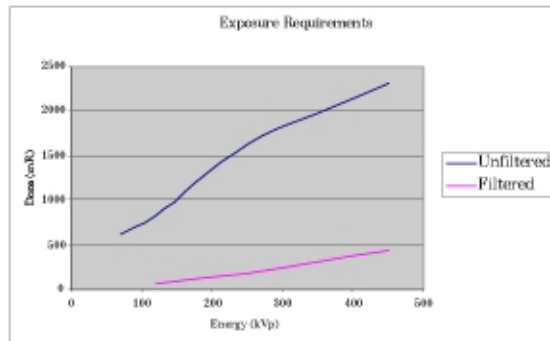


Modulation transfer functions were determined for 4 different screens at 3 different energies



KD\_CASIS #13

Exposure requirements (ER)\* were determined for x-ray source energies 70 to 450 kVp

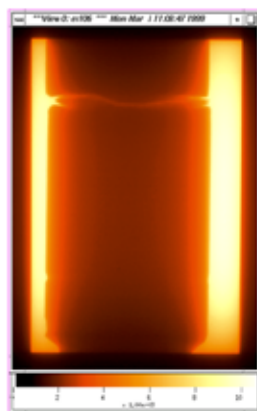


\* ER is exposure that produces near saturation (light field) image

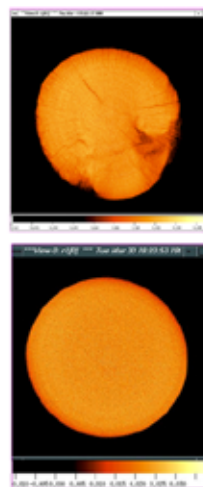
KD\_CASIS #14



a-Si computed tomography of steel bridge pin  
3 inch diameter at 9 mV



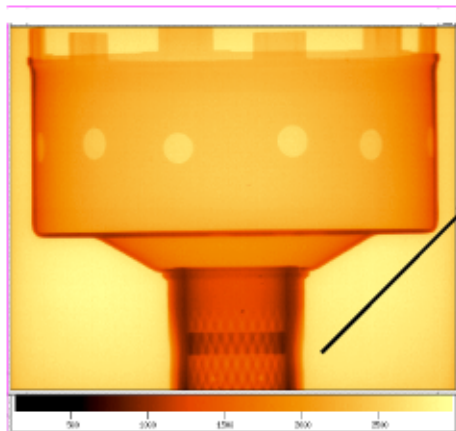
Digital Radiograph



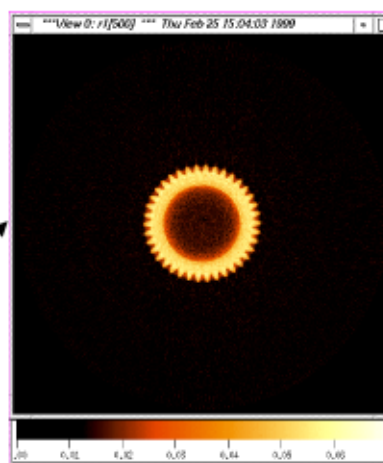
Tomograms

KD\_CASIS #15

a-Si computed tomography of prototype clutch  
retainer gear at 4 mV



Digital Radiograph



Tomogram

KD\_CASIS #16



## Conclusions

---



A-Si flat panel imager provides a practical new imaging method for radiology and computed tomography, and opens new possibilities for fast digital image acquisition and quantitative analysis.

KD\_CASIS #17







---

# Astronomical Imaging

---



## **Speckle Imaging of Saturn's Moon Titan**

**Seran Gibbard, Bruce Macintosh, Claire Max, Donald Gavel: LLNL**  
**Imke de Pater, Henry Roe: UC Berkeley**  
**Andrea Ghez: UCLA**  
**Christopher McKay: NASA Ames**  
**Eliot Young: Southwest Research Institute**

## **OUTLINE**

- 1. Speckle Imaging**
- 2. Titan background**
- 3. Our Titan observations**
- 4. Modelling Titan's surface and atmosphere**
- 5. Titan's surface**
- 6. Conclusions**



## **STEPS IN SPECKLE PROCESSING**

- **Take short-exposure images (shorter than time constant for turbulence)**
- **Remove bias, sky, flat-field effects on each frame**
- **Combine image Fourier transforms to get object's power spectrum**
- **Then use bispectrum technique (Lohmann et al. 1983) to estimate phase**
- **Then convolve with power spectrum of unresolved point source, re-convolve with theoretical telescope PSF**
- **Resolution is limited by diffraction limit of telescope, not by 'seeing'**
- **Diffraction-limited resolution for 2 microns at Keck is 0.04 arcseconds**
- **Technique limited to bright objects since each short exposure must have sufficient signal/noise**

## **TITAN**

- **Largest moon of Saturn – larger than Mercury**
- **Substantial nitrogen/methane atmosphere**
- **Photolysis of methane – atmosphere has complex organic chemistry**
- **Hydrocarbon haze may 'rain' out on surface**
- **Possible hydrocarbon 'seas'?**
- **Analogue for early Earth before life**



## SPECKLE IMAGING OF TITAN

- **Dates:** Sept. 6 1996, Oct. 10-13 1997, July 29-Aug 1 1998, Oct. 7-14 1998 (UT)
- **Telescope:** 10-m W.M. Keck
- **Instrument:** Near Infrared Camera (NIRC) with image converter
- **Plate scale:** 0.02 arcseconds per pixel
- **Resolution (diffraction limited)** 0.04" at K
- **Wavelengths:**    K' (1.96-2.29  $\mu\text{m}$ , 4360-5100/cm)  
                              H (1.49-1.82  $\mu\text{m}$ , 5495-6710/cm)
- **Data taken in sets of 100 frames, each frame 150 or 200 ms of integration**

## ATMOSPHERIC MODEL

- **About 60% of light received from Titan at K' and 70% of light at H comes from atmosphere rather than surface**
- **By analyzing this light can constrain atmosphere**
- **Atmospheric model must reproduce:**
  - Observed limb brightening
  - Observed asymmetry in N/S brightness (S hemisphere currently brighter)
  - Overall brightness of planet
  - Don't get negative surface albedo!
- **Adjust haze parameters to minimize difference between model and limb of planet**
- **Model gives light received from the atmosphere; the remainder of the light comes from the surface**



## **WHAT IS TITAN'S SURFACE MADE OF?**

- **Bright regions may be ice, ice/rock mixture**
- **Dark regions could be:**
  - **Crater basins filled with impact or volcanic material**
  - **Low-lying regions into which organics are washed by methane rain in bright highlands**
  - **Areas of infrared-dark ice**
  - **Mixture of ice and organics?**
  - **'Seas' of liquid hydrocarbons**

## **CONCLUSIONS**

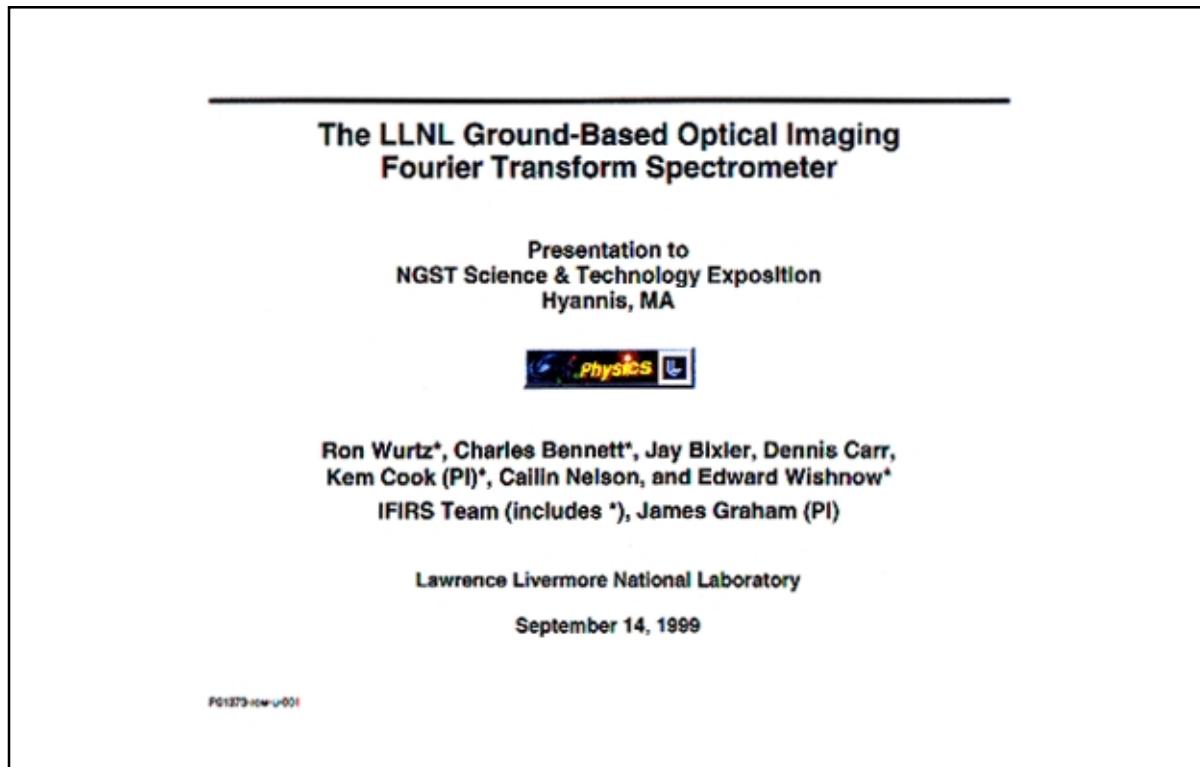
- 1. Speckle imaging gives excellent resolution for bright solar system objects**
- 2. Limited to broadband filters**
- 3. Atmosphere models useful to separate Titan's surface from atmosphere**
- 4. Surface albedo map of Titan reveals bright and dark areas—possible solid organics or liquid hydrocarbons**
- 5. Adaptive optics will allow narrowband imaging of Titan to better separate surface and atmosphere**



# Astronomical Imaging Fourier Transform Spectroscopy: Technology and Techniques

*Ron Wurtz*

Our group is developing an astronomical ground-based facility-class visible/IR imaging Fourier transform spectrograph (IFTS). Beginning with a run at Kitt Peak in March 1999, this project has been demonstrating astronomical IFTS in the low flux regime with bench-top technology and techniques. Our demonstrations include multi-object spectroscopy, simultaneously obtained narrow-band imaging, time-induced noise removal, dispersed and Fourier transform hybrid spectroscopy, etc. We will discuss these techniques as well as the techniques of data calibration, reduction, and extraction.



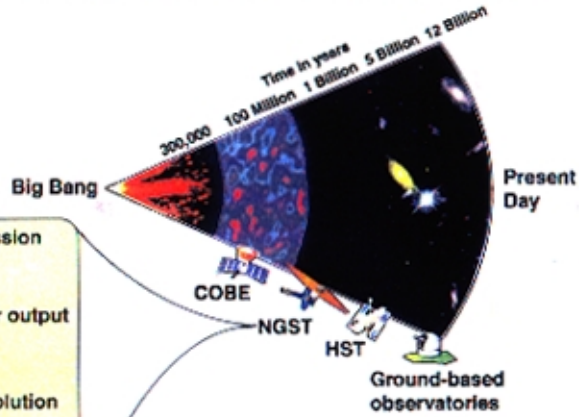


## The Next Generation Space Telescope (NGST) will probe "the Dark Zone" and answer key questions

- How and when were the first stars and galaxies born?
- What is the mysterious dark matter?
- What is the shape of the universe?

A faster, cheaper, better mission  
~\$1 billion launch 2007

- 1-5  $\mu\text{m}$ 
  - Redshifted stellar output
- 8-m diffraction limited
  - Faint objects
  - High angular resolution
- 50 K operating temperature
  - Low background
- L2 orbit
  - Low background

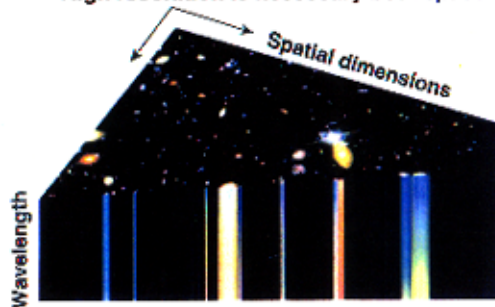


F01254-10w-016



## With unprecedented sensitivity comes unprecedented instrumental requirements

- In very deep images, virtually every pixel contains interesting sources
- Spectroscopy for every pixel is needed for scientific understanding
- High resolution is necessary both spatially and spectrally



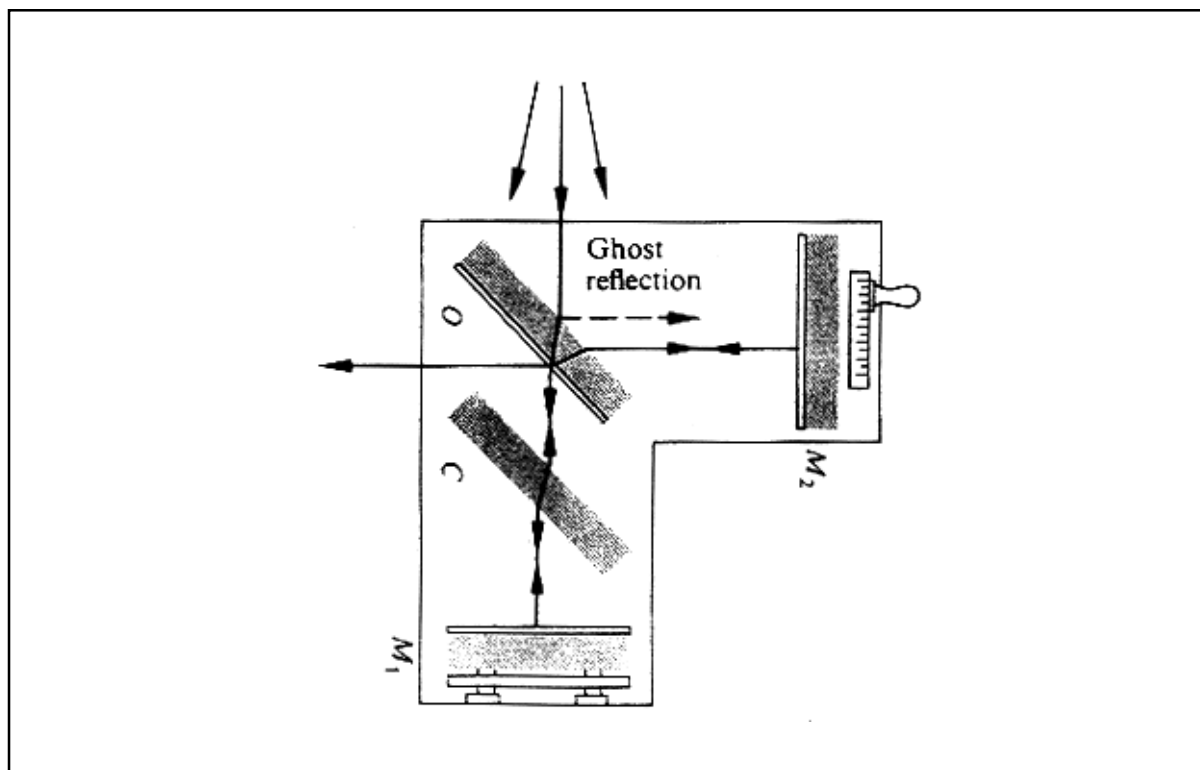
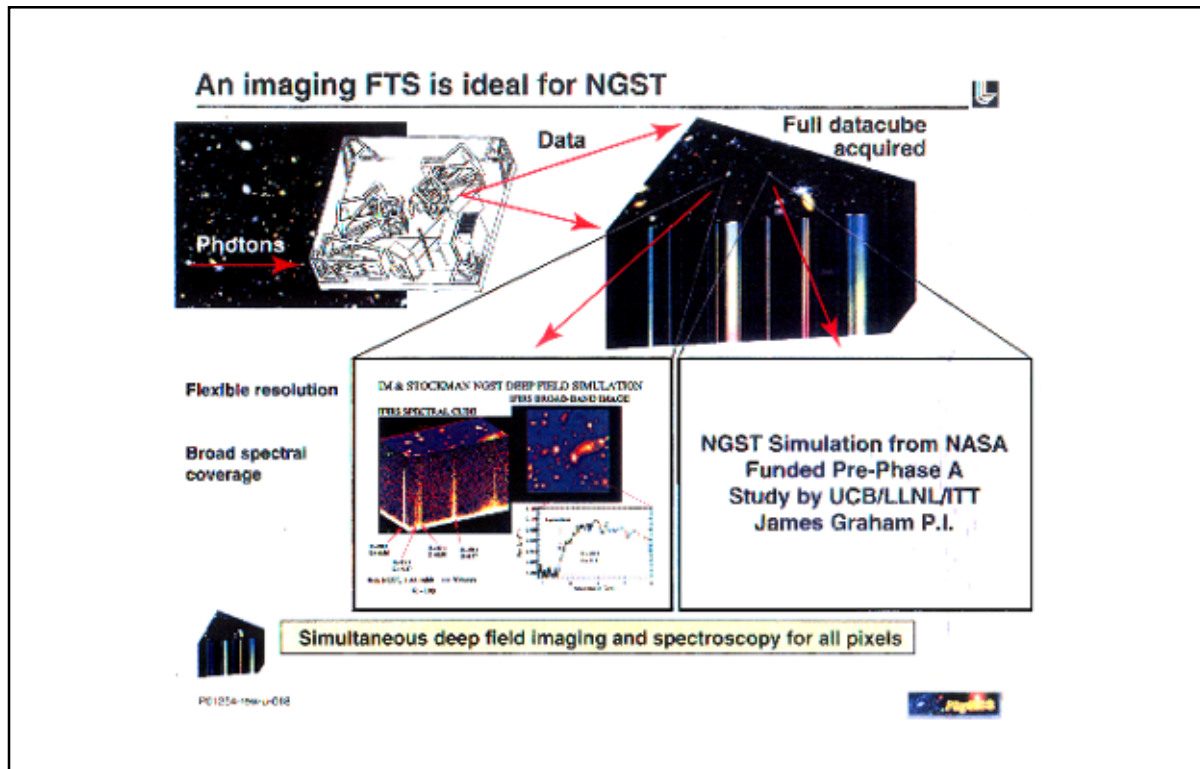
The "Hubble Deep Field" contains thousands of galaxies in a supposedly "empty" field

LLNL is an acknowledged world leader in infrared imaging spectroscopy:  
Developed HIRIS technology and hosted 3-D Imaging Conference

F01254-10w-017

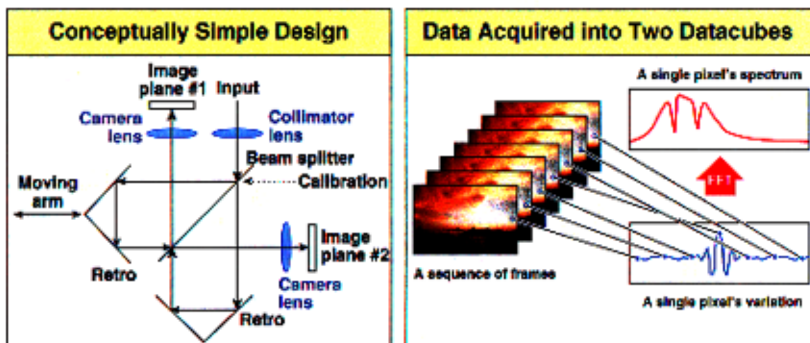








## Review of Imaging Fourier Transform Spectrometers



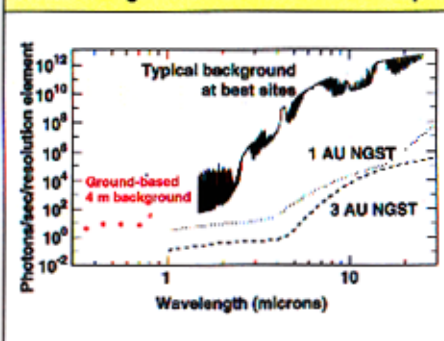
P01375-000-002



## NGST IFTS Concept Tests are Possible with Ground-based Optical IFTS



**Ground-Based Visible Observing Environment  
Is a Parallel to NGST NIR/MIR. Visible is  
a Natural Region to Test Instrument Concepts**



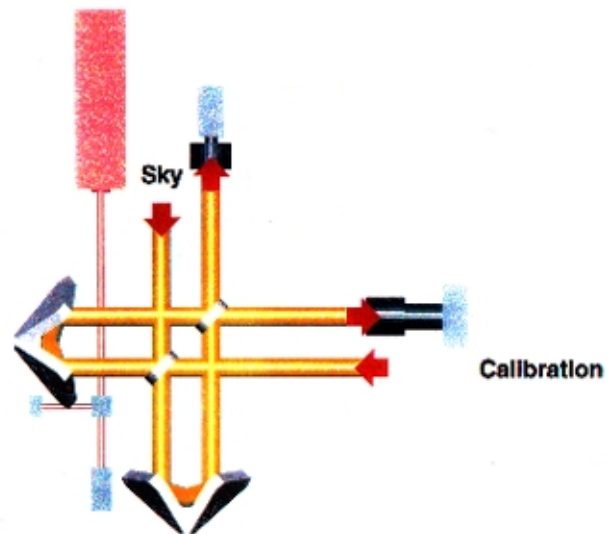
- Build Optical Astronomical Imaging Four-Port FTS and demonstrate:
  - Visible FTS tolerances are met with existing technology
  - A new regime: long dwell time step scan (<1 Hz)
  - Two output advantage
  - Hybrid dispersed FTS
  - Data reduction

P01375-000-003





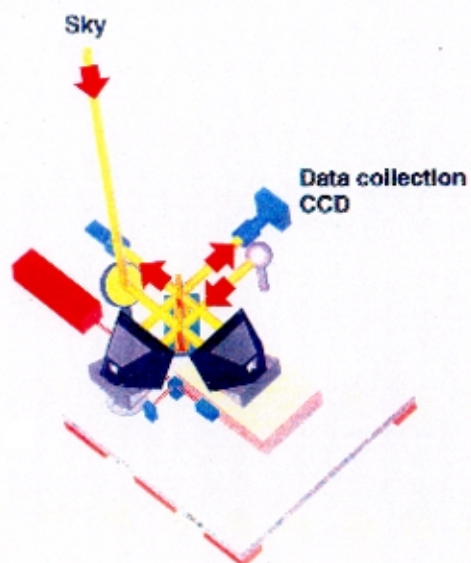
### Tabletop IFTS with coaligned metrology system



P01254-10W-U-003



### IFTS layout at NSO





## McMath-Pierce Main Spectrograph Port

- 1.5 Meter Primary
- Friendly to Visitor Instruments
- Horizontally Mounted Field De-rotator



Incoming Vertical Beam

Rotating Optical Bench (interferometer goes here)

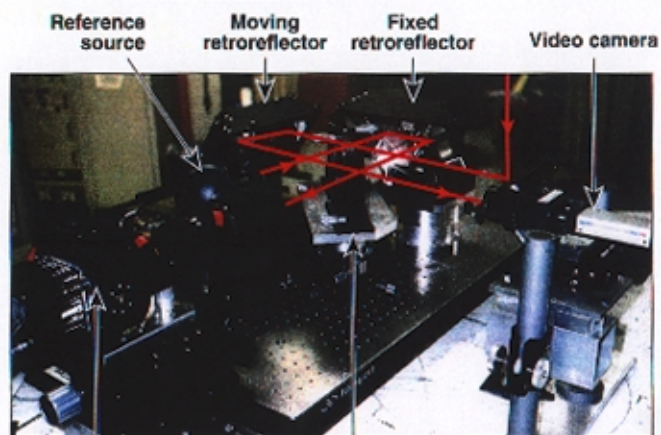
Rotating Spectrograph



UCB/LLNL/STScI/JHU/NRL/ITT

4

## The LLNL Testbed Interferometer



1024x1024 TE-cooled  
CCD camera

Beamsplitters

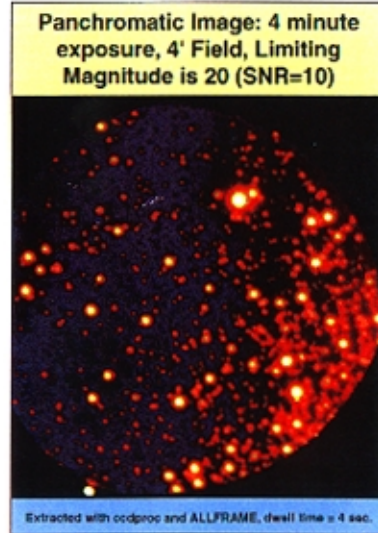
- First Generation: LLNL Instrument is Open Design
- Second Generation: "Facility"-like Instrument

P01373-10w-u-004



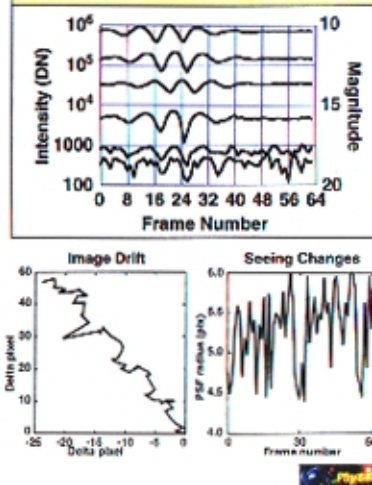


### Kitt Peak: Southern edge of M4 (Single Output Port)



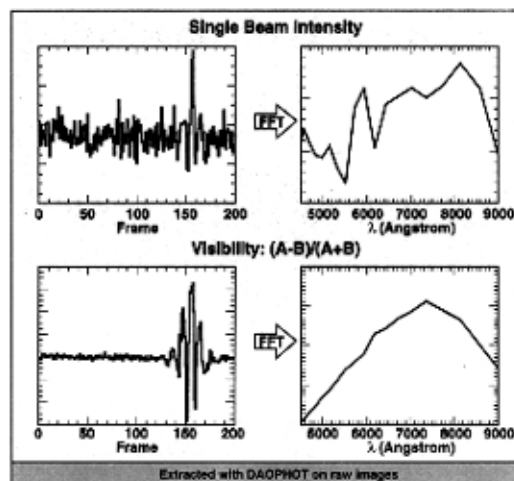
P01251-raw-u-005

#### Representative Interferograms For Magnitudes 10 to 18



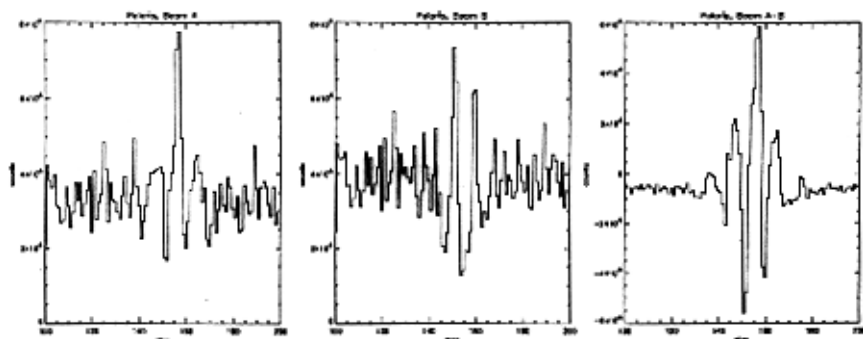
### Polaris: Demonstration of advantage of two output ports

Kitt Peak data were obtained with one science output port. Adding a second science output allows for common mode rejection reduces noise from internal and external time variations in source and intervening absorbers.



P01273-raw-u-007

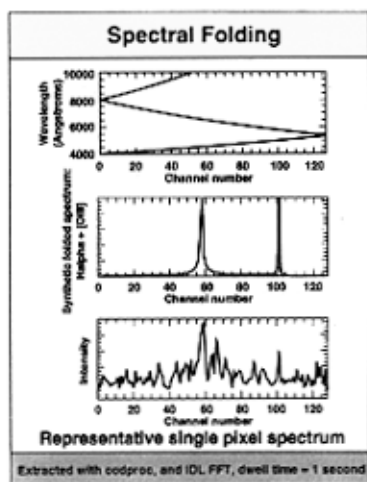




### Kitt Peak: Orion Nebula (Single Output Port)



Extended region with widely spaced emission lines



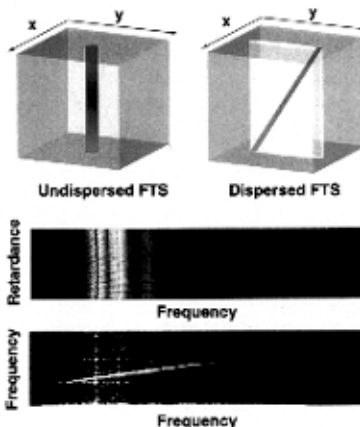
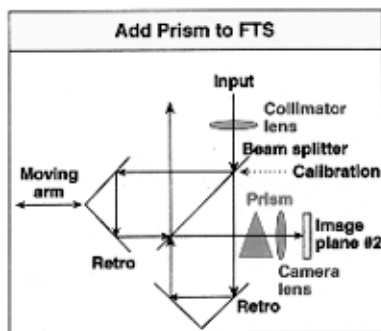
PG 1273-mw-uc06





## Polaris: Hybrid FTS and dispersing spectrometer (single output port)

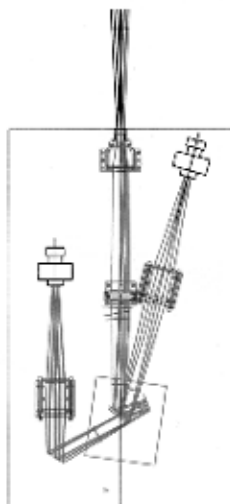
Inserting a Dispersing Element (prism) just before the camera makes a Hybrid Dispersed FTS



P01873-new-u-008



## Visible imaging FTS: NGST technology testbed



- Benchtop visible IFTS has probed NGST regime:
  - Shorter wavelengths (tight tolerances)
  - Long integration times (step scan)
  - Technology readiness
  - Data reduction technique
  - Varying spectral resolution
  - Spectral folding
  - Common mode rejection
  - Hybrid FTS plus disperser
- Next:
  - Visible Astronomical IFTS using technology developed specifically for NGST
  - Fielded at 4-10m class telescope
    - Fall 1999: ARC 3.5 m
    - 2000: Gemini/Keck?

P01375-new-u-009





## "First Light" Science Results from the Telescope Adaptive Optics System

Claire Max  
Institute of Geophysics and Planetary Physics  
and University Relations Program  
LLNL



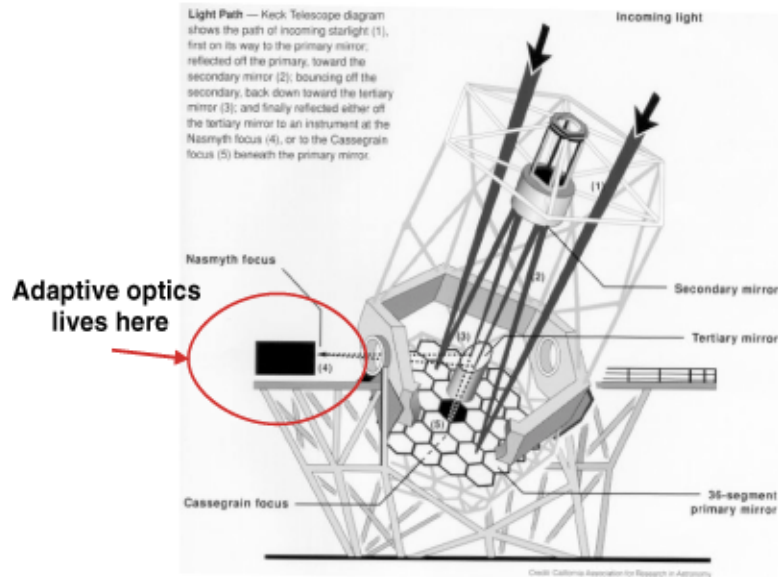
Adaptive optics system has been a joint  
project between Keck and LLNL staff



- Keck people :
  - P. Wizinowich , S. Acton, O. Lai, C. Shelton, P. Stomski , ...
- LLNL people (laser program, engineering dept, IGPP):
  - C. Max (PI), J. Brase , D. Gavel, B. Macintosh, S. Olivier, J. An,  
H. Friedman, G. Erbert , K. Waltjen , K. Avicola , K. Watson, ...



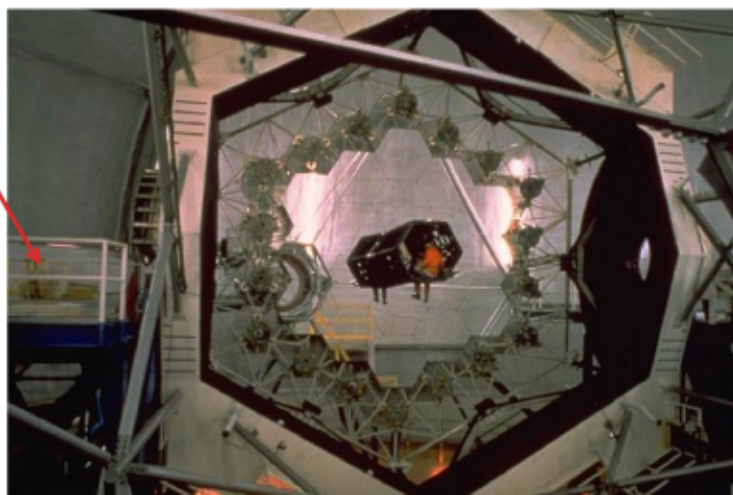
## The Keck Telescope



## Keck primary mirror before final ring of segments was installed



**Nasmyth platform**





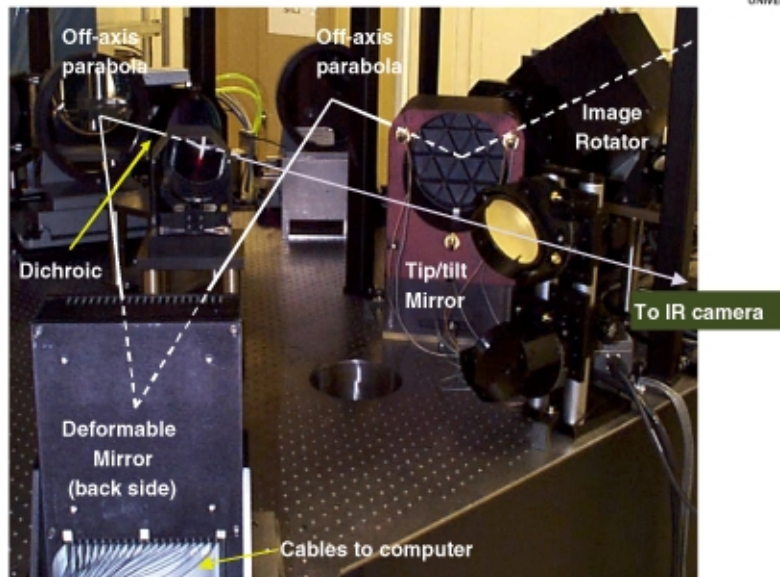
## Nasmyth platform enclosure : a meat locker!



- Temperature-controlled via glycol cooling (some say it's a morgue)
- Rule in the dome: anything emitting more than 100W has to be glycol-cooled
- Reason: "dome seeing" has been responsible for much of the poor image quality in past



## Path of light to science camera



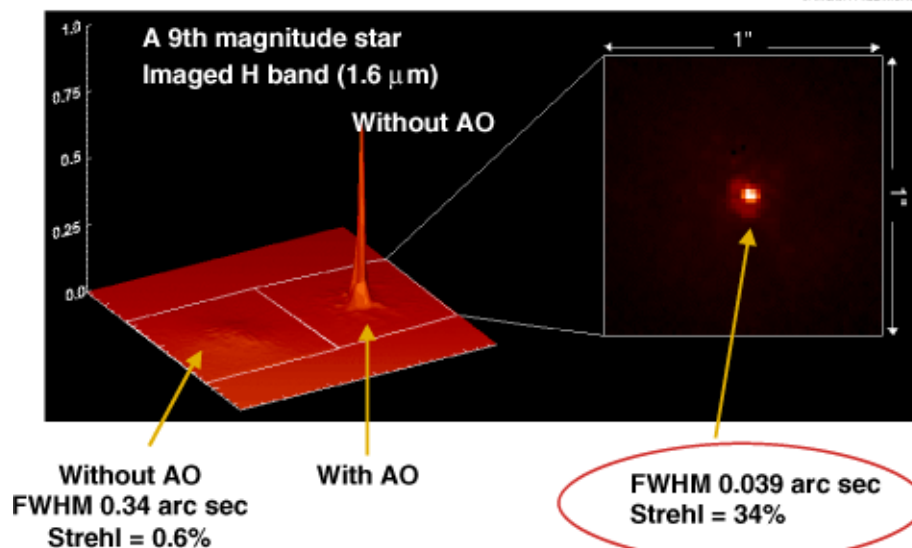


## **“First Light” science results**



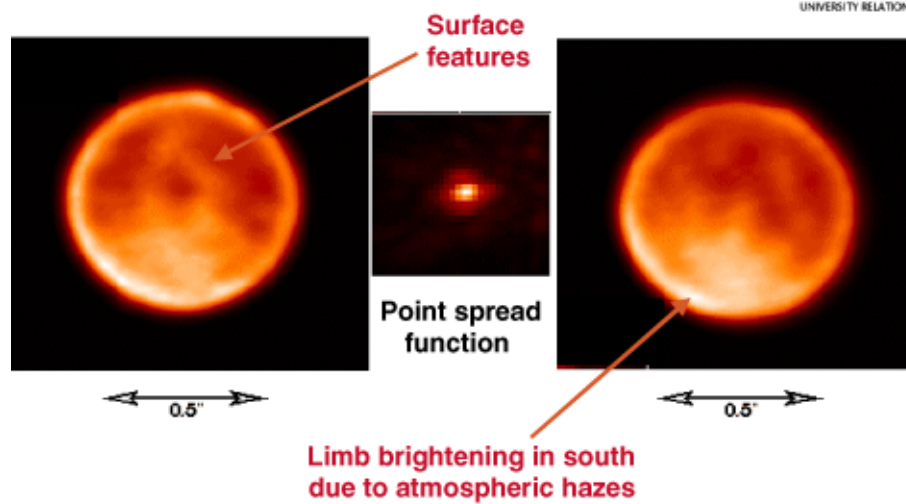
- Performance of AO system on bright stars
- Detail on the surface of Saturn’s moon Titan
- Spectacular images of storms on Neptune
- Motion of stars within 0.3 parsecs of the black hole at the center of our Galaxy

## **System performance on bright stars is spectacular!**



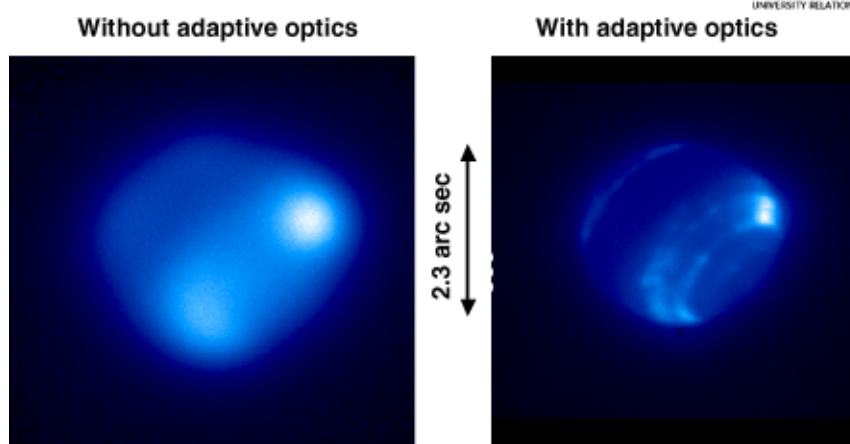


## Keck images of Titan at two different phases ( $1.6\ \mu\text{m}$ )



S. Gibbard, B. Macintosh, C. Max, I. de Pater (UCB)

## Neptune at 1.6 microns



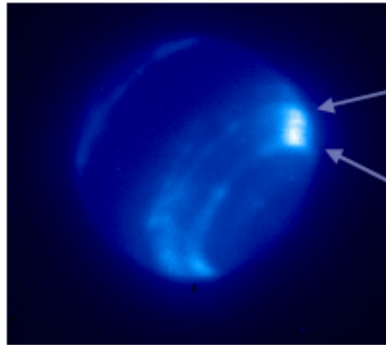
S. Gibbard, B. Macintosh, C. Max, I. de Pater (UCB)



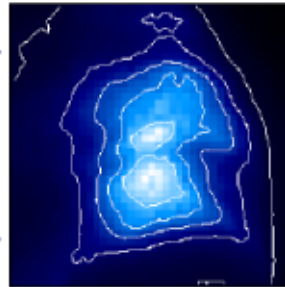
## Details of Neptune's bright storm at a scale of 400 - 500 km



Square root color map

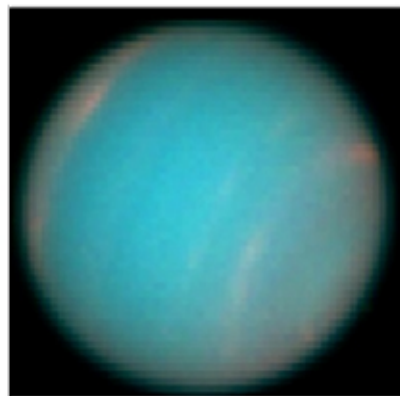


Linear color map

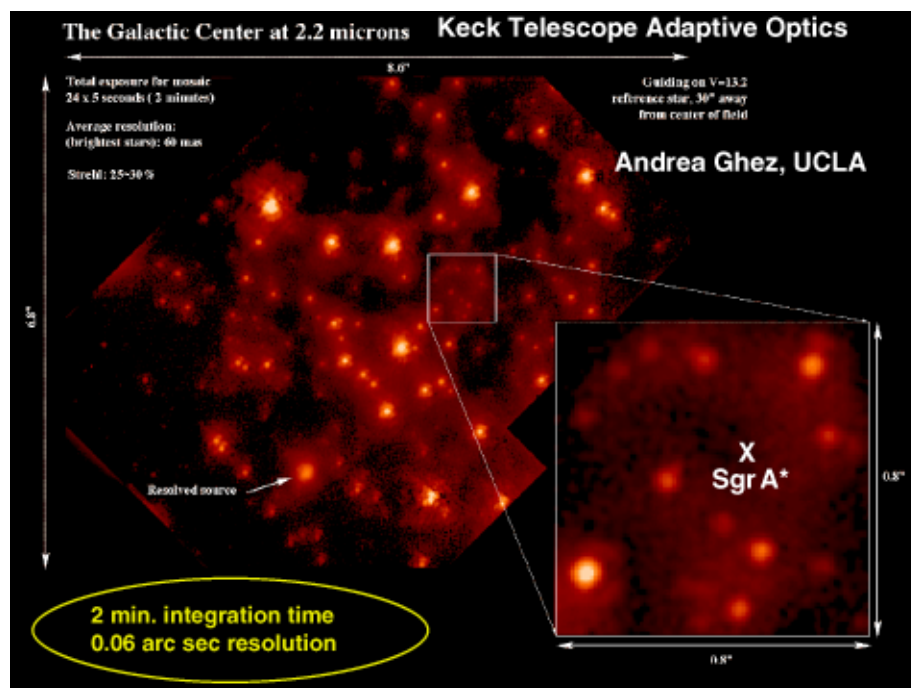
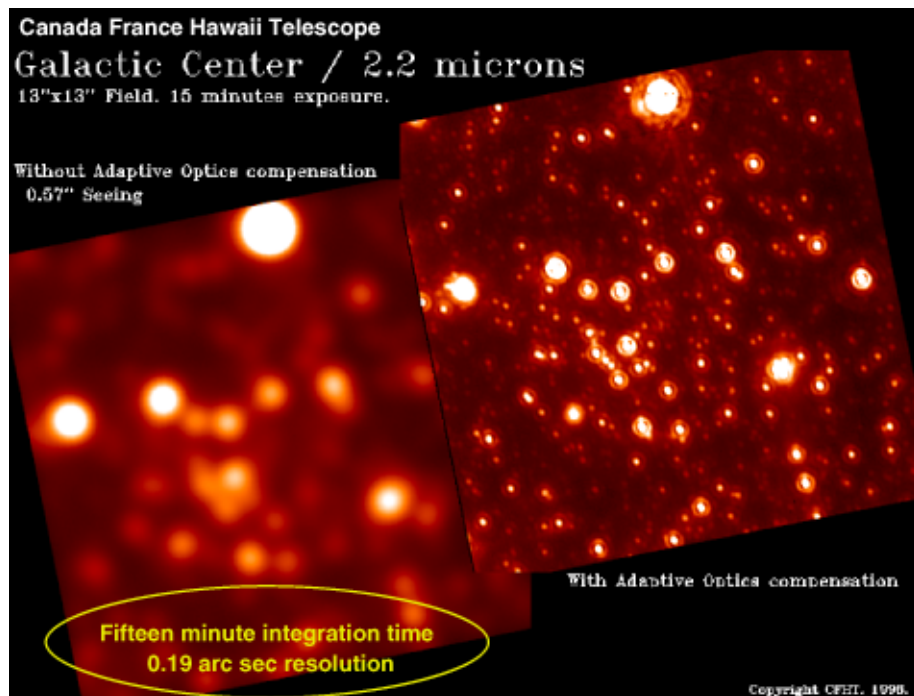


Each pixel is 0.017 arc sec  
 $\Delta x = 375$  km at Neptune

## Hubble Space Telescope image of Neptune









## Summary



- Keck AO system is the most ambitious of the new generation of large telescopes (8 - 10 m diameter mirrors)
- Keck AO performance is truly outstanding:
  - Without AO:        FWHM ~ 0.6 arc sec
  - With AO:            FWHM ~ 0.04 arc sec (!)
- Better than Hubble Space Telescope
  - Spatial resolution 4 X better than HST in infra-red
  - 10 m telescope diameter  $\Rightarrow$  can detect fainter sources
  - Spectroscopy will be spectacular

**This really is the dawn of a new age for astronomy!**

## Coda: Overview of new NSF Center for Adaptive Optics



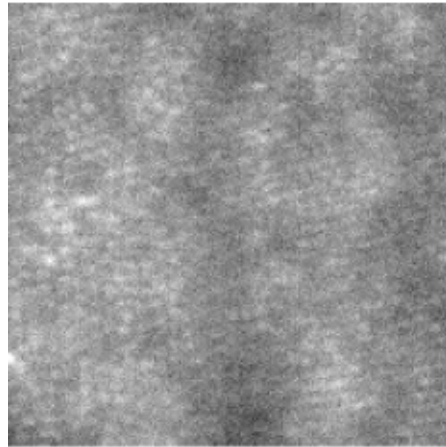
- NSF Science and Technology Centers:
  - Multidisciplinary, education & outreach, industrial partners
  - Competition last year: several hundred applied, 5 selected
- Center for Adaptive Optics:
  - Lead by UC Santa Cruz - home of UC Observatory and Lick
  - Jerry Nelson PI (inventor of Keck Telescope)
  - Very strong participation by LLNL
  - Many other institutions are partners

**Adaptive optics for astronomy and vision science**

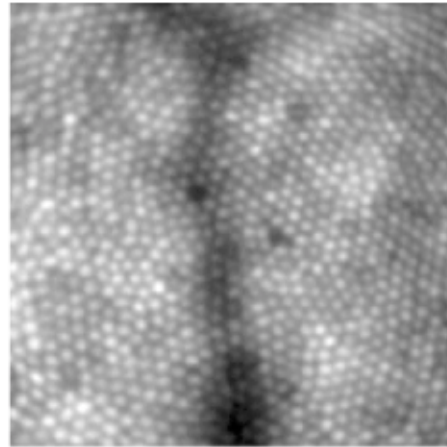


## Adaptive Optics Provides the Highest Resolution Images of the Living Retina

> 3x increase in transverse resolution

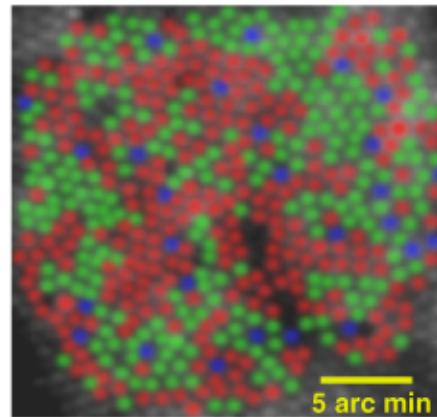
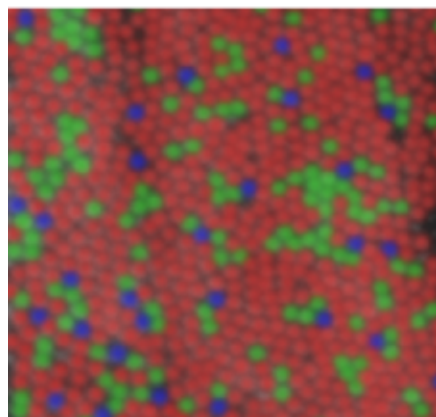


Without AO



With AO: resolve individual cones

## Human Color Vision: huge variability from one individual to the next!



First images of the human trichromatic cone mosaic, Roorda and Williams, Nature, 1999. Note how different the two cone patterns are!



# Searches for Low Mass Extra-solar Planets

*Jian Ge & Dave Erskine, LLNL*

One of the most exciting fields in astronomy is the search for extra-solar planets by measuring the extremely small Doppler velocity shifts on starlight caused by planets. However, two main factors are preventing current planet searches from detecting low mass extra-solar planets such as Saturn-like and Earth-like planets. One factor is related to the unstable and irregular point spread functions with the current seeing-limited echelle spectrographs. The other is related to the intrinsic turbulence and oscillations in the star's photosphere. These two factors limit the radial velocity resolution beyond  $\sim 3$  m/s (Marcy & Butler 1998).

The novel spectrally dispersive interferometer, or fringing spectrometer (Erskine & Ge 1998; 1999), developed by us at LLNL, provides one solution to the current problems. The simple and stable instrument responses enables significant reduction of radial velocity error to less than  $\sim 1$  m/s. The laboratory-based testing with this instrument has demonstrated a 0.76 m/s resolution capability, more than three times better than the current state-of-the-art. The instrument is being coupled to the Leuschner 30inch telescope for planet searches— similar velocity resolution for starlight is expected soon. Once candidate stars showing a few m/s velocity variation are identified with this instrument, further observations with another LLNL instrument, which is a novel adaptive optics very high resolution spectrograph, will follow up to monitor their stellar photosphere activity. This will help the detection of low mass extra-solar planets. The combination of the dispersive interferometer and the adaptive optics very high resolution spectrograph will be the most promising way for low mass extra-solar planet searches in the next decade.



**Presentation  
at 99 CASIS  
Workshop at  
LLNL**



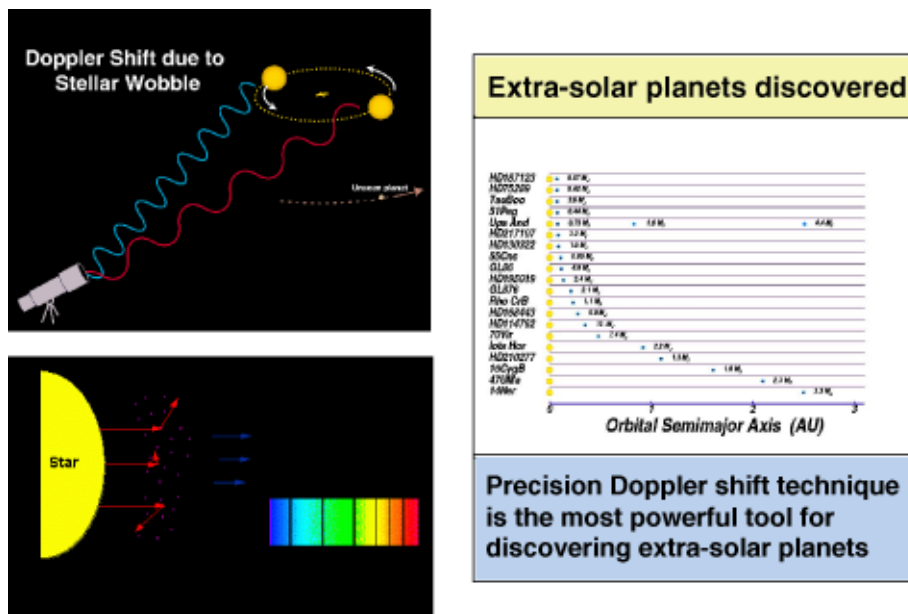
**UCRL IL10168**

## Searches for Low Mass Extra-solar Planets

Nov. 12, 1999

**Jian Ge & Dave Erskine**  
Lawrence Livermore National Laboratory

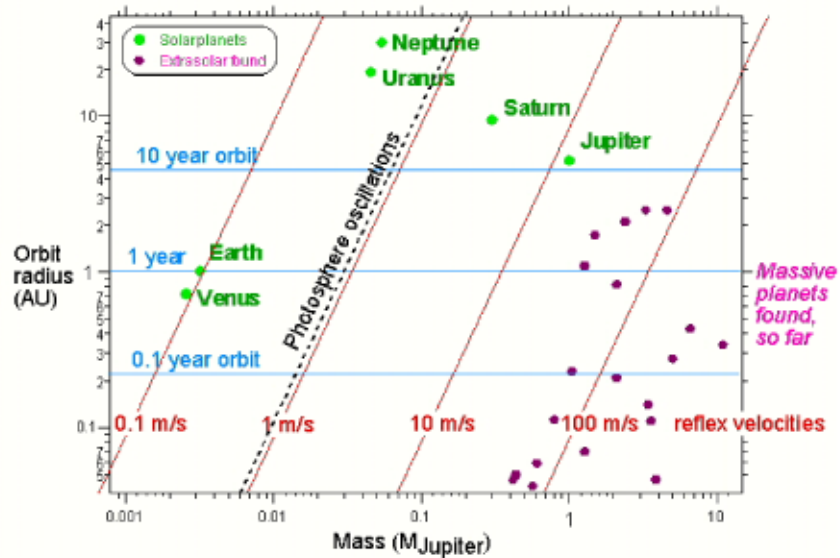
**Collaborators:**  
Gibor Basri(UCB), Bruce Macintosh (IGPP), Mike Rushford (LLNL), Mike Lloyd-Hart, Roger Angel (UA)





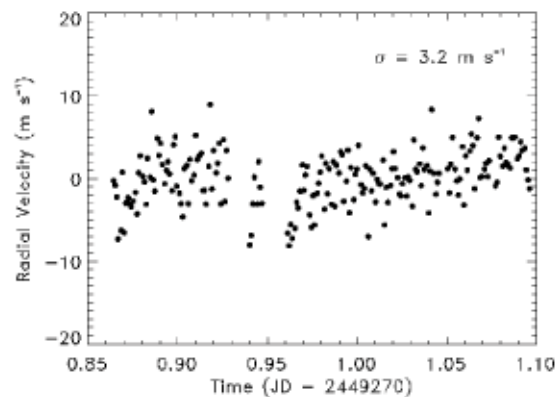
### Are there any solar-like extrasolar planets?

Need to search for smaller velocity signatures.



Doppler velocity precision < 3 m/s in order to detect solar-like extra-solar planets

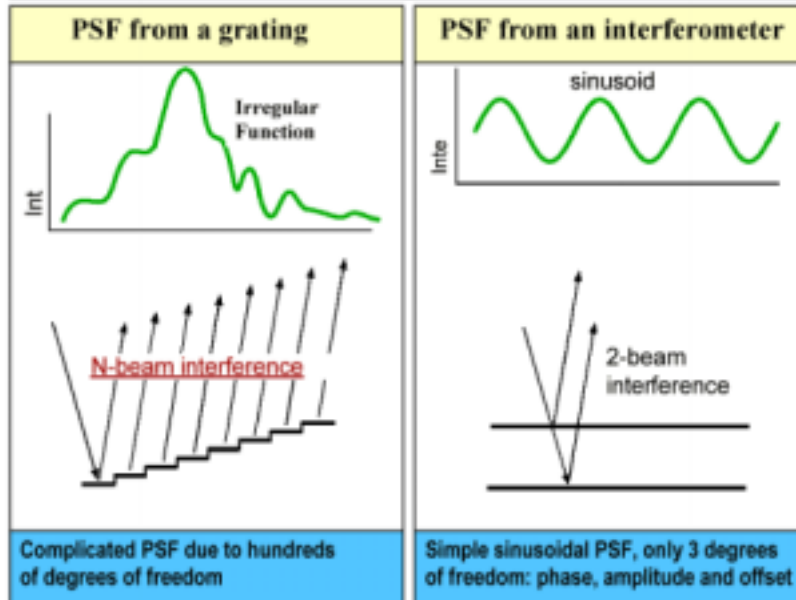
### Current State-of-the-art Doppler Radial Velocity Precision



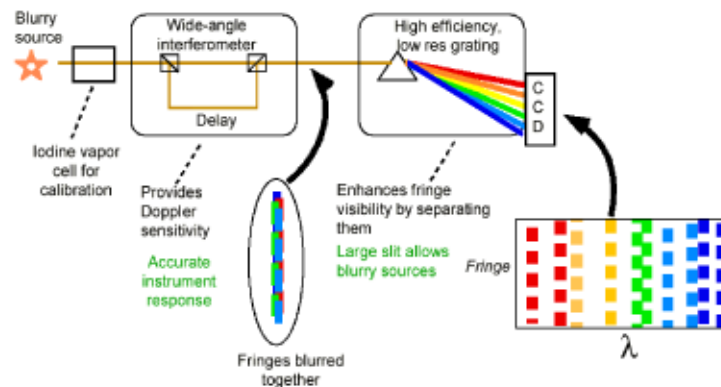
Current radial velocity precision barely allows detection of Jupiter-like planets, not Saturn-like & Earth-like planets



The simplest instrument response of an interferometer offers more precise measurements of slight Doppler shifts



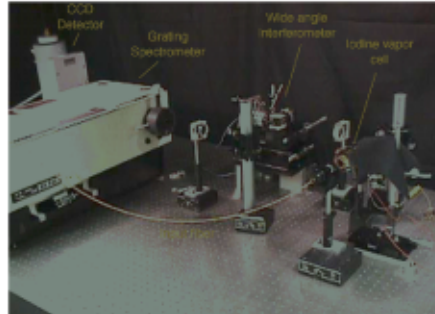
Dispersive Interferometer (or fringing spectrometer) combine best features of interferometer and grating spectrometer



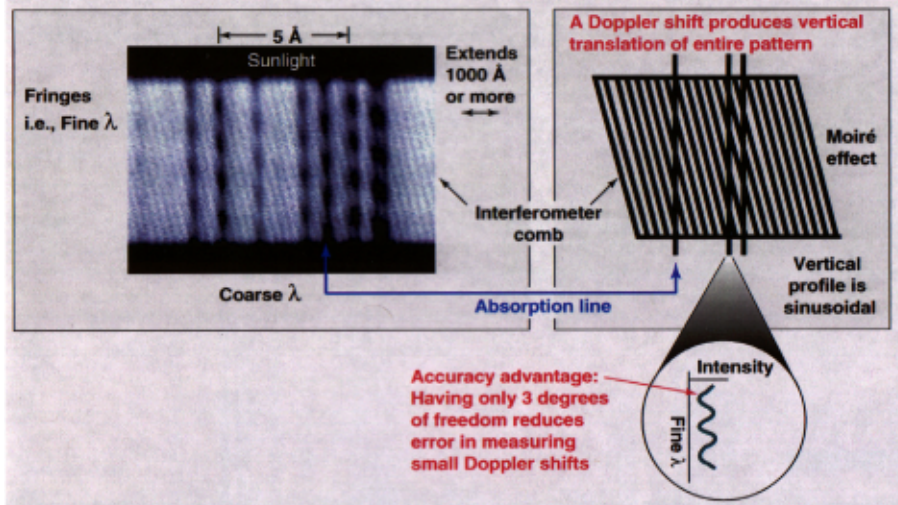
- The interferometer produces pure and simple sinusoidal fringes (PSFs)
- The spectrometer disperses white fringes into different color channels to increase fringe visibility for precision phase measurements



### Prototype dispersive interferometer for initial lab testing



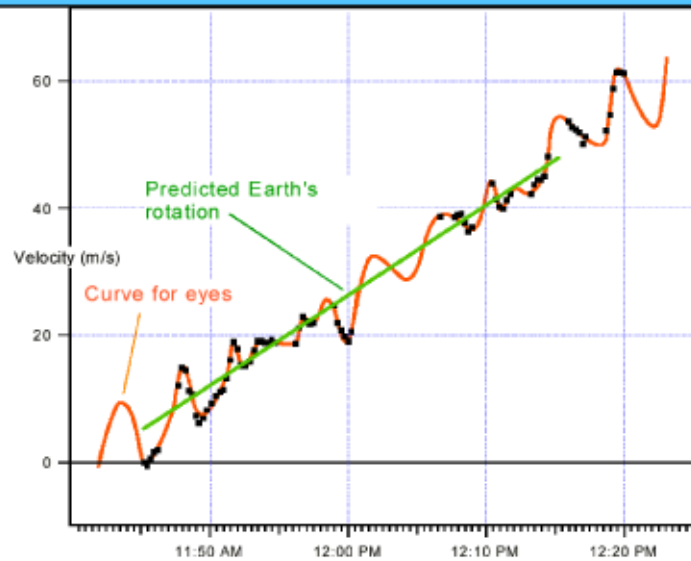
### Solar fringing spectrum obtained by the prototype dispersive interferometer



Solar absorption line Moiré fringes are used to measure tiny phase shifts caused by radial velocity variations

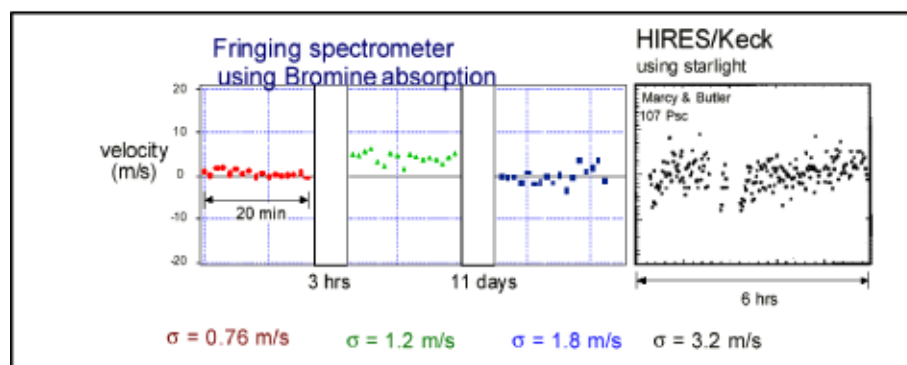


**Observations of the Earth's Diurnal Motion and Solar 5 minutes' Oscillation with the Prototype**



We need further observations to confirm the solar oscillation detection due to the unstable heliostat setup during the measurements

**Null tests: Prototype's radial velocity precision is better than current state-of-the-art**

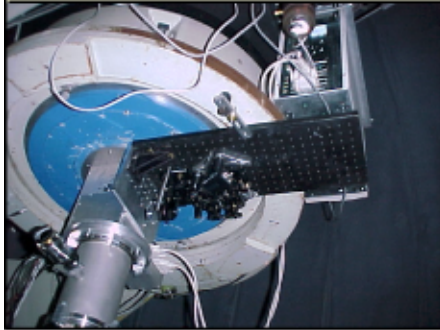


- Short term radial velocity precision already 3x better
- Additional controls on interferometer cavity etc. will reduce systematic errors for better long term precision



**A new fiber-fed dispersive interferometer is ready for starlight testing**

Fiber-fed assembly at the Leuschner 30 inch telescope for testing, 10/1/99



Lick 40 inch telescope will be used for first light stellar observations

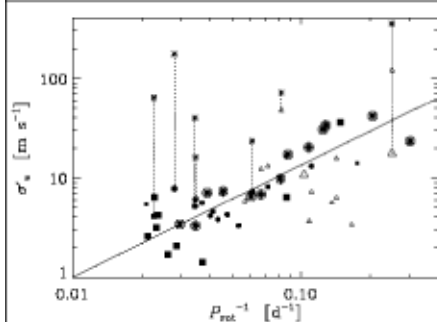


**A survey for low mass extra-solar planets will start at the Lick 40 inch in December 1999**

**Stellar activity-related radial velocity variations limit Doppler precision for low mass planet searches**

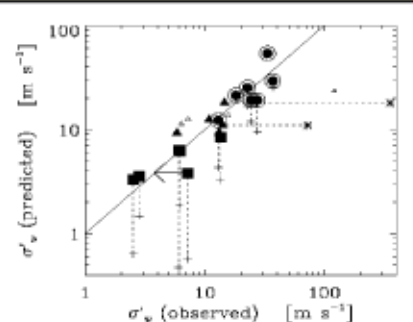


Radial velocity variation,  $\sigma_v$ , vs. stellar rotational period,  $P_{\text{rot}}$  (Saar et al. 1998)



Old, slowly rotating stars have the lowest activity-related radial velocity noise

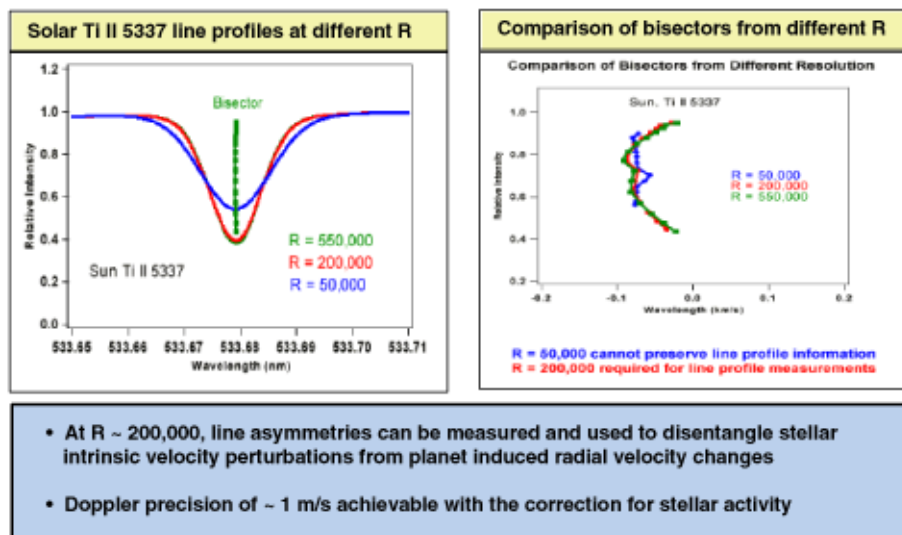
The observed radial velocity variation vs. theoretical prediction (Saar et al. 1998)



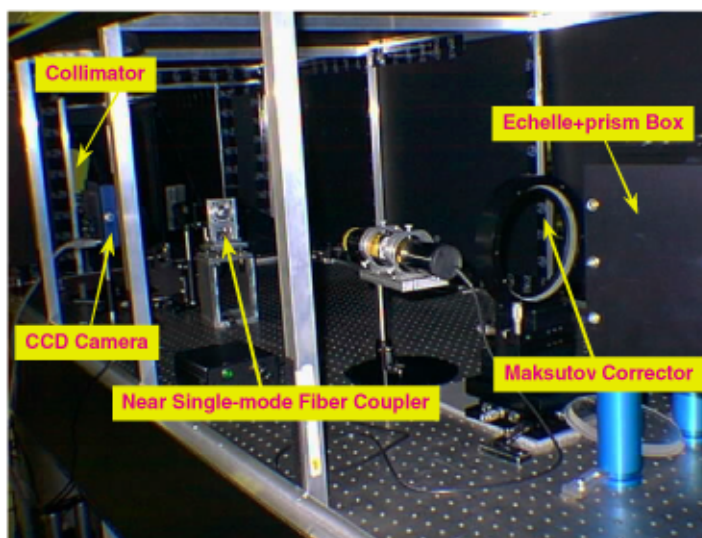
Convection dominates velocity variations in F stars vs. starspot activity in G, K stars



# Study of stellar absorption line profile is the key to understand different contribution to stellar radial velocity variation

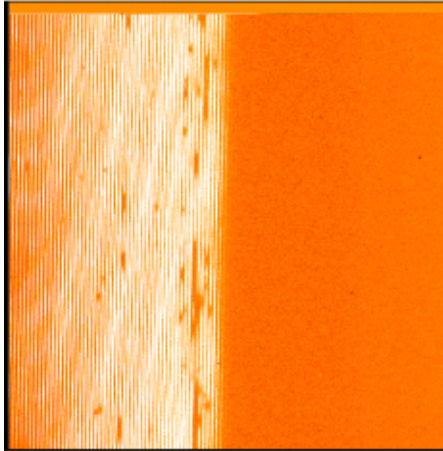


## Setup of a New AO Optical Cross-dispersed Echelle Spectrograph at Steward Observatory, Nov. 1998, $R = 200,000$ , $0.35\text{-}1.0\ \mu\text{m}$ (Ge et al.)



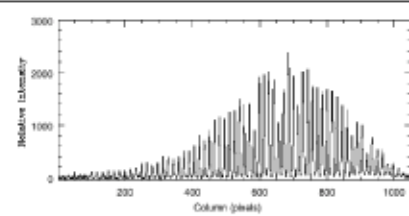


**Solar Spectrum Obtained with the AO Spectrograph and 2Kx4K CCD Camera at Steward Obs., May 99 (Ge et al.1999)**



Part of solar spectrum covered by the 2Kx4K CCD camera

- ~ 100 orders covered by the 2Kx4K CCD
- ~ 4000 Å coverage per exposure
- Minimum order separation is 240  $\mu\text{m}$
- Total detection efficiency at the SOR 3.5m is ~ 3%
- S/N ~ 200, 10 min for V = 7 mag.



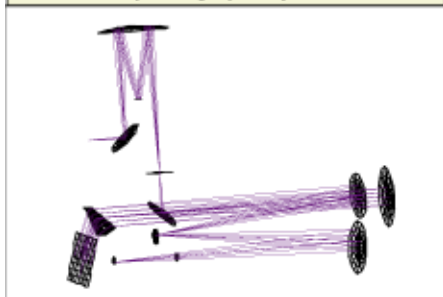
Cross-dispersed orders, order separation > 16 pixels

The spectrograph will be commissioned at the SOR 3.5m telescope in Dec. 99

**A compact IR immersion echelle spectrograph with adaptive optics (LISPEC) is being developed at LLNL to achieve  $R = 100,000\text{-}300,000$  in  $1.2\text{-}5.5 \mu\text{m}$  (FY99-FY02)**

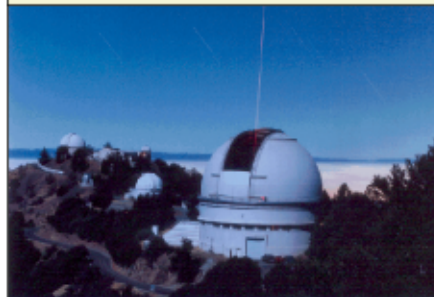


Spectrograph Layout



- Silicon immersion grating being developed at the Microfabrication center
- IR spectrograph and its prototype being developed at IGPP

Scientific Observations



- The spectrograph will be coupled with adaptive optics at the Lick for study of stellar magnetic fields
- IR spectroscopy a factor of 2 or more sensitive to magnetic Zeeman broadening than optical one



# Demonstration of a Diffractive Telescope

*Sham Dixit*

*Co-authors:*

*Mike Rushford, Ian Barton, Frank Patterson, Leslie Summers, John Prior,  
Holly Cagle, Jerry Britten, Katy Lu, Mike Perry and Rod Hyde*

We present results of the imaging and color correction properties of a telescope that uses diffractive fresnel lenses. The theory behind a color correction of a diffractive telescope will be discussed. This will be followed by a description of the fabrication of the Fresnel lenses and the fielding of the telescope. Images of the planets in the solar system taken using the telescope will be presented.

---

## Demonstration of a diffractive telescope

---



Sham Dixit

Mike Rushford , Ian Barton, Frank Patterson, Leslie Summers, John Prior  
Holly Cagle, Jerry Britten , Katy Lu, Mike Perry and Rod Hyde

Presentation to:  
CISIS Workshop  
November 11, 1999



## Eyeglass is a new type of space telescope



### Contains two cooperating spacecraft

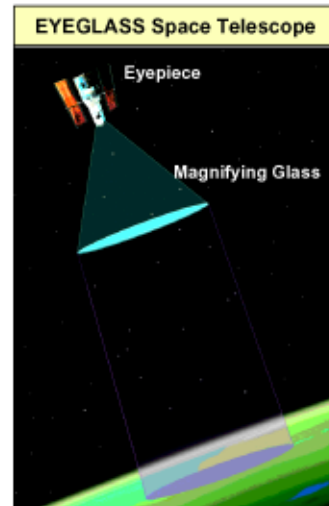
- Separated by a few kilometers
- Tied together via micro-gee propulsion

### Magnifying Glass: Thin diffractive lens

- Provides the large aperture
- Lightweight and easy-to-deploy
- With very loose surface tolerances

### Eyepiece: Compact, meter-class, space telescope

- Moves along focal surface, collecting light
- This is a conventional, low-risk, vehicle

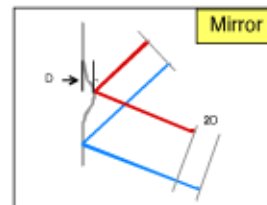


## Why Do Things This Way?



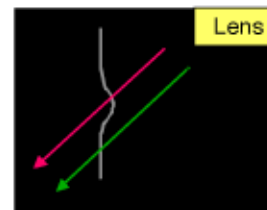
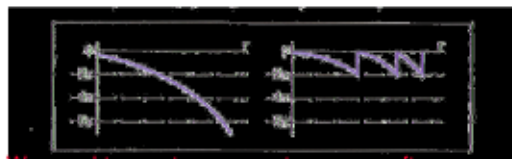
### Thin lenses have much looser tolerances than mirrors

- Surface ripples on a mirror are optically doubled
- Ripples on a thin lens are optically cancelled
  - This effect is best for weak, low F#, lenses
- Surface tolerances are ~ 1 cm, not ~ 500 angstroms
  - An F/100 lens is 160,000 times better than a mirror



### We must use diffractive lenses, not refractive ones

- A conventional lens is too thick and heavy
- A diffractive lens is very much thinner and can be made lightweight and foldable

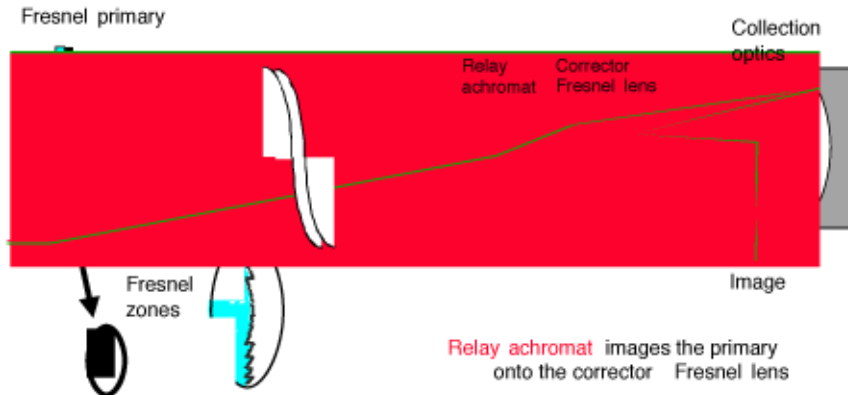


### We need to use two separate spacecraft

- Focal length is kilometers
  - Product of large aperture (for resolution) and high F# (for tolerance gains)
- These lengths are practical in space, by splitting the telescope into two parts



## Color correction in a diffractive telescope



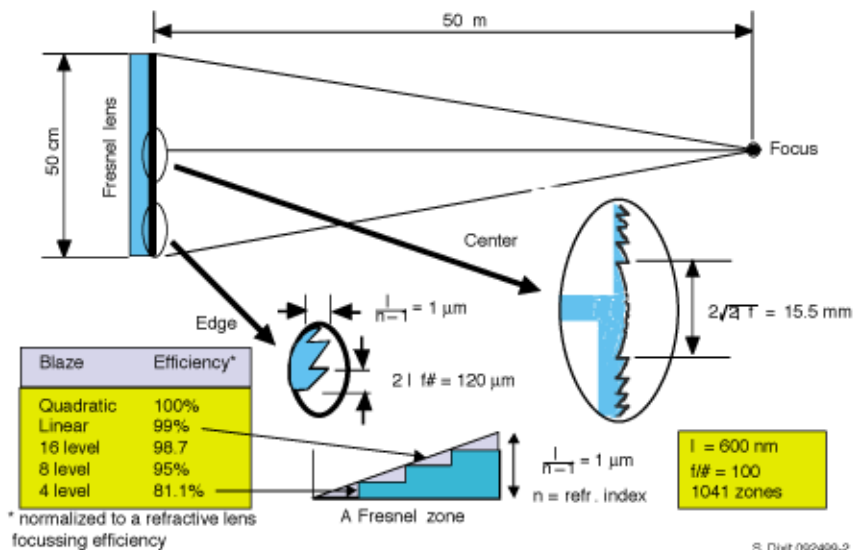
Features:  
Large aperture (25-50 m)  
Long  $f\#$   
Fabricated on a thin membrane

SND - 990804-1

## Characteristics of a diffractive Fresnel lens

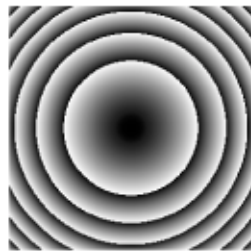


Example shown is for a 50-cm diameter  $f/100$  lens





The Fresnel lens phase profile is fabricated by overlaying multiple binary mask patterns in a lithographic process



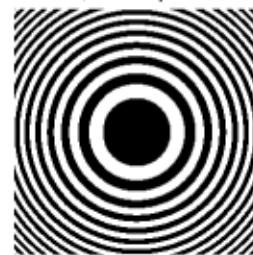
Phase profile (mod  $2\pi$ )  
phase step depth =  $t$

To achieve high diffraction efficiency, we need:

- high precision binary masks ( $<1\ \mu\text{m}$  accuracy)
- good alignment ( $<1\ \mu\text{m}$  precision)
- precise etch depth control (10 -20 nm)



mask 1; etch depth  $t/2$



mask 2; etch depth  $t/4$

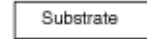
Coat with photoresist

Substrate

Expose with UV

Develop & etch

Wash

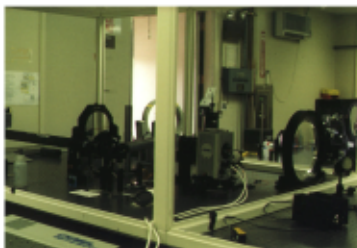
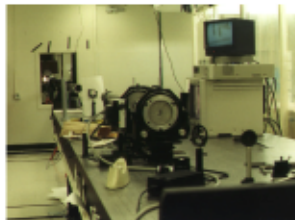


S. Dixit 002499-3

## 20 cm Diffractive Telescope: Hardware



Full System



Beam Expander and Diffractive Primary



Fresnel Corrector and Schwarzschild Collector



## Color Correction Has Been Demonstrated

LDRD 20 cm Telescope

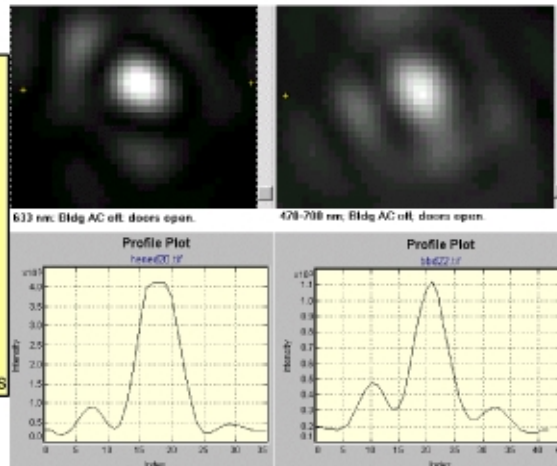
Delivers color-corrected spot

Same size for broad-band  
as for monochromatic

Diffraction-limited spot-size

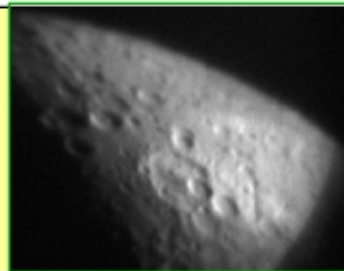
Predicted Airy -diameter: 96 $\mu$ m

Same as measured: 96 pixels

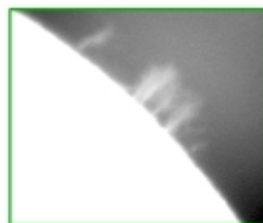


## Diffraction Telescope Has Taken Astronomical Images

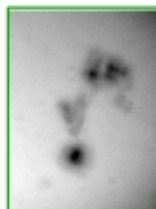
- Despite complications
  - Telescope is immobile, with narrow FOV
  - Image must be piped-in with turning mirrors
- We have obtained clear astronomical images
  - White light: Moon, Jupiter, Saturn, Sun-spots
  - Narrow band: Solar flares



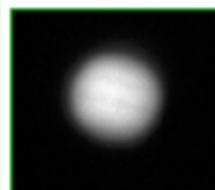
Lunar Surface



Solar Flares



Sun-Spots



Jupiter



Saturn

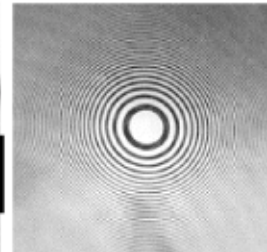


We have successfully fabricated the binary mask for the 50-cm Fresnel lens



Pictured:  
Mike Rushford

Expanded view of the center of the mask



Mask parameters:

Aperture diameter = 50 cm  
Focal length = 50 m  
Design wavelength = 0.6  $\mu$ m  
  
Central zone dia. = 11 mm  
Last zone width = 60  $\mu$ m  
Number of zones = 1041

S. Dixit 002499-5

Nova laser bay is being used to field the 50-cm diffractive telescope

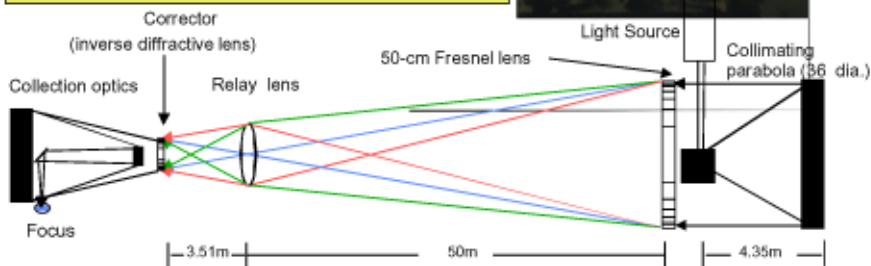
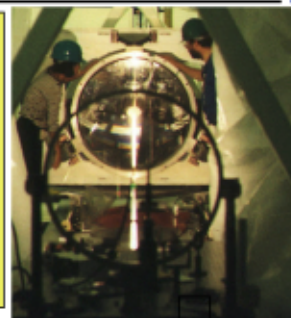


Telescope characteristics:

50-cm diameter primary Fresnel lens  
50 m focal length  
color correction using a Fresnel lens  
(~1" diameter; written by P. Maker JPL)

Use the laser space frame for Beamline 10 on the Nova laser

67 m (220 ft) straight path  
Temperature stability ~1 degree  
Vibration isolated space frame  
Clean room environment  
Real estate and other hardware available from Nova



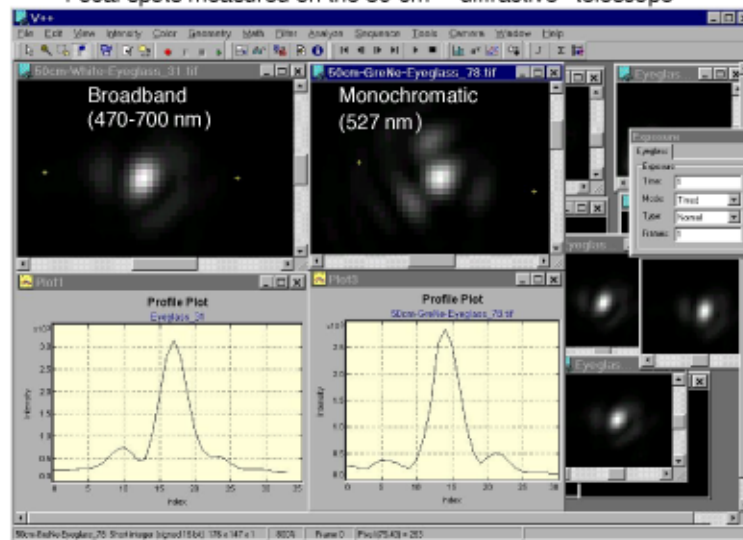
S. Dixit 002499-9



We have demonstrated near-diffraction limited imaging on the 50-cm diffractive telescope in the Nova laser bay



Focal spots measured on the 50-cm diffractive telescope



## Summary



- discussed how diffractive telescopes offer a powerful alternative way for broad-band earth and space observation
  - require large Fresnel lens on thin membrane as the primary can be fully color corrected
- demonstrated diffraction limited performance for the 20-cm and 50-cm telescopes
- obtained astronomical images using the 20-cm telescope







---

# Image Processing

---



## Detection of Vertical Obstructions in SAR Images

UCRL-VG-136063

*Sailes K. Sengupta and Ursula Goldstein*

*Paper submitted to the Signal and Imaging Sciences Workshop  
November 11-12, 1999*

### *Abstract*

Finding vertical obstructions such as towers, buildings, hanging wires, etc. is an important part of generating up-to-date maps or related spatial databases. SAR imagery can be a useful source of information in this process because cloud cover does not interfere with imaging and because many of these objects produce strong radar returns. The problem of automatically detecting vertical obstructions has been addressed in this work using methods derived from pattern recognition theory and morphological image processing. Hanging wires approximated by low curvature parabolas have been detected in this work by a modified version of the Randomized Hough Transform (RHT). Towers, buildings and other 'bright' or 'dark' objects have been detected by clustering pixels in a feature space consisting of neighborhood statistics of pixels. Towers are localized from among the bright objects using their shape features. Possible locations of buildings have been detected using the morphological features of their shadows obtained from clustering. Finally, the need for the creation of an appropriate perceptual information framework has been indicated in situations where a successful integration of information from different sensors and existing models is essential.

Work performed under the auspices of the U.S. Department of Energy by Lawrence Livermore National Laboratory under Contract W-7405-ENG-48.



## Use of Morphological Operators and Pattern Recognition Techniques for Sorting the FIRST Data

*Deanne Proctor*

The application of morphological and pattern recognition techniques to the sorting of FIRST data is presented. FIRST (Faint Images of the Radio Sky at Twenty Centimeters) is designed to produce the radio equivalent of the Palomar Observatory Sky Survey, using the National Radio Astronomical Observatory Very Large Array. A catalog of source positions and Gaussian-fit parameters for the sources is available.

The goal of this part of the project is the automatic classification of types of sources, since it is expected that there will be on the order of one million sources in the final catalog.

The data consists of radio maps, each approximately  $1500 \times 1100$  pixels and containing on average about 30 sources per map. The resolution of these maps is  $5''$  (approximately 3 pixels). The sources range in size from a few pixels to 70 or 80 pixels maximum diameter and have various morphologies. Of particular interest is the extraction of 'bent doubles', core jets, and gravitational lens candidates. These interesting types constitute on the order of percent of the sources.

The emphasis in this presentation is on image-based pattern recognition. We report on the use of morphological operators for source extraction and compare use of decision trees and neural networks in the classification of source morphologies and accounting for the finite resolution.

Keywords: pattern recognition, morphology



USE OF MORPHOLOGICAL OPERATORS  
AND  
PATTERN RECOGNITION TECHNIQUES  
FOR  
SORTING THE FIRST DATA

Deanne Proctor  
Scientific Computing Applications, Computations  
Lawrence Livermore National Laboratory

November, 1999.

RESOURCES

- FIRST maps  
(FIRST survey, Robert Becker, P.I.)
- IDL Interactive Data Language  
(Research Systems, Inc.)
- IDL UIT database software  
(Landsman)
- IDL FIRST catalog in database form  
(Michael Gregg)
- OC1 decision tree software  
(Murthy, Salzberg, Kasif, Johns Hopkins Univ.)
- SNNS Stuttgart Neural Network Simulator  
(Institute for Parallel and Distributed High Performance Systems, Univ. of Stuttgart)

Thanks to IGPP, Charles Alcock, Kem Cook for office  
and computing facilities.



## OUTLINE

## I. INTRODUCTION

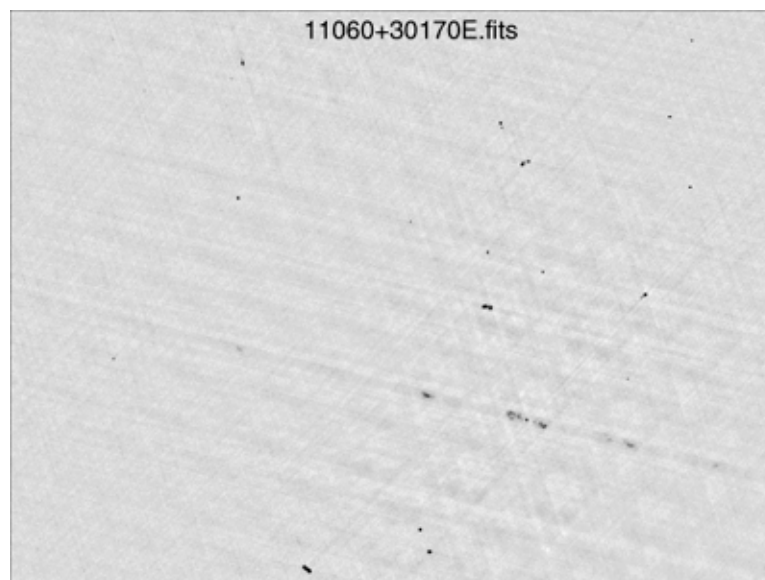
## II. SOURCE EXTRACTION

- A. Catalog
- B. Morphological Operators
- C. Groupings

## III. PATTERN RECOGNITION

- A. Invariance
- B. Training and Test Sets
- C. Feature Selection
- D. Decision Trees and Neural Networks

## IV. RESULTS





## SOURCE EXTRACTION

## • CATALOG

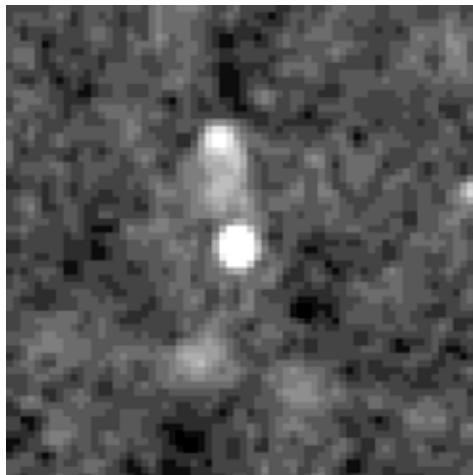
Up to 4 Gaussians fit per source island.

Software to group catalog sources by proximity.

Singles	71%
Doubles	18%
Triples	6%
Four and more	5%

## • MORPHOLOGICAL

Erode, Dilate and Label Region



n98\_109888.first.fits

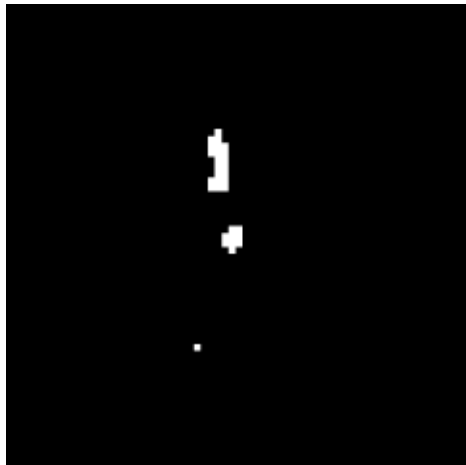
image





n98\_109888.first.fits

3 sigma threshold of image



n98\_109888.first.fits

erode(image\_thr3,s5)



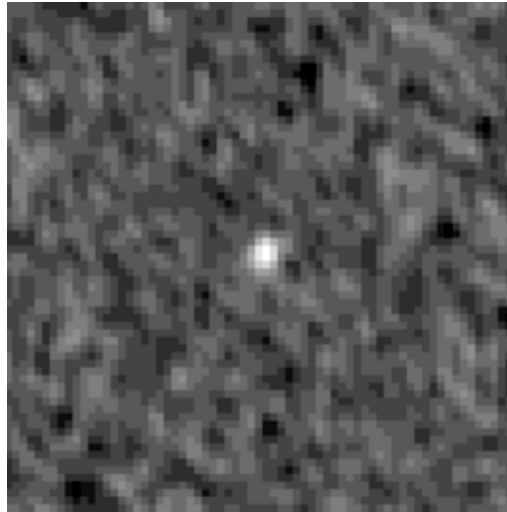


n98\_109888.first.fits

dilate(erode(image\_thr3,s5),s5)

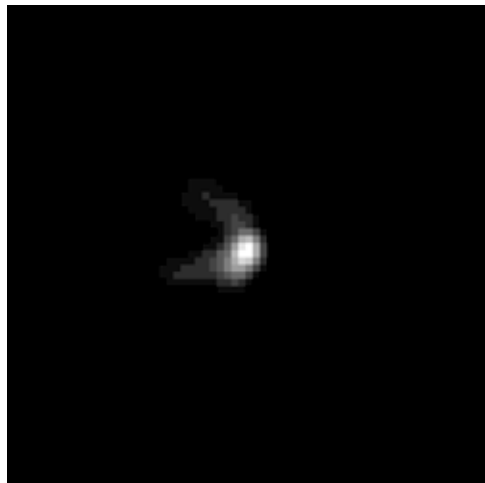






s98\_000744.fits

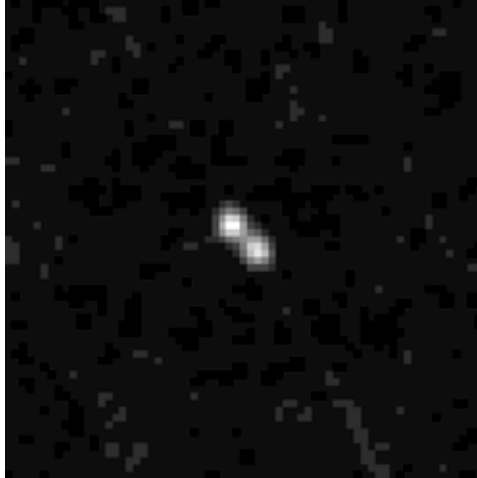
SINGLE



s98\_00210.first.fits

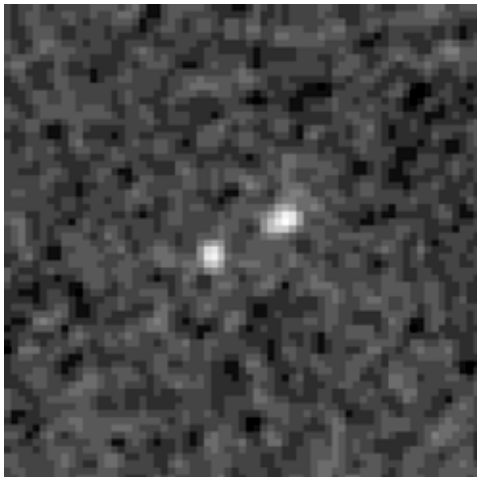
SINGLE





north98\_074343.fits

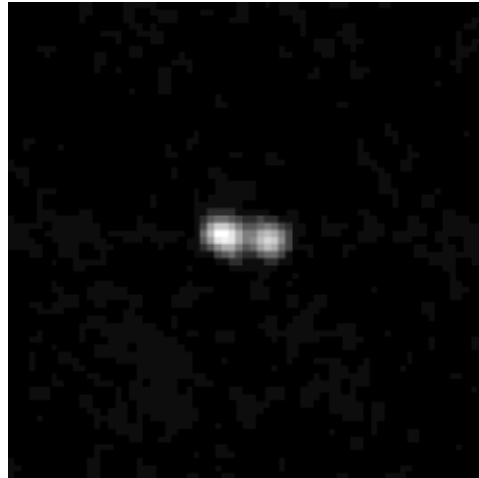
DOUBLE



north98\_076235.fits

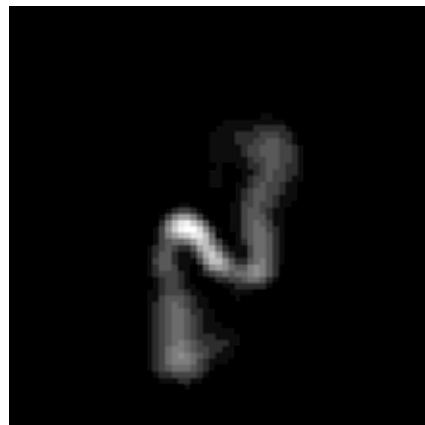
DOUBLE





north98\_103126.fits

DOUBLE



north98\_069940\_069971.fits

SPIRAL





south98\_008442\_008447.fits

NAT

### CLASSICAL INVARIANCES REQUIRED

- Translation
- Rotation
- Scale - this is the problem,  
due to finite point spread function



## TRAINING SET - Visual Classifications

233 Bent  
 426 Somewhat Bent  
 1255 Non-Bent  
 721 ? (too noisy to tell)  
 10 Misc.

2645 Total

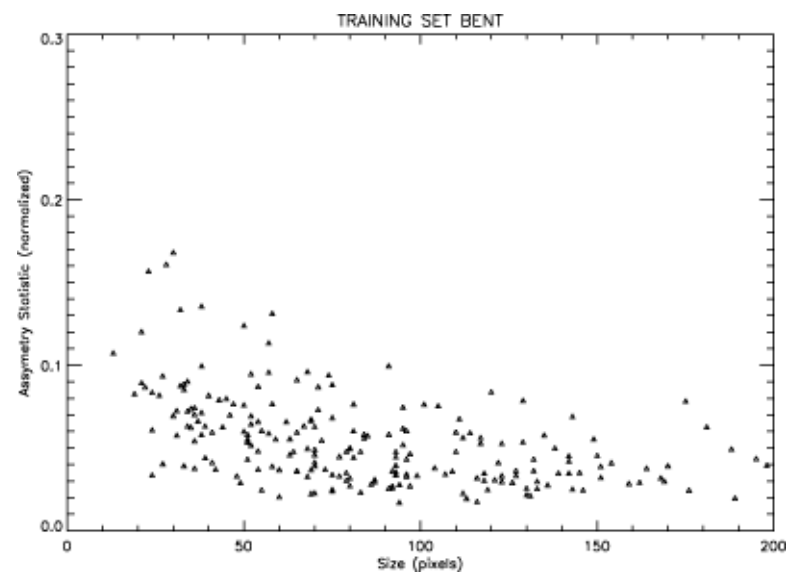
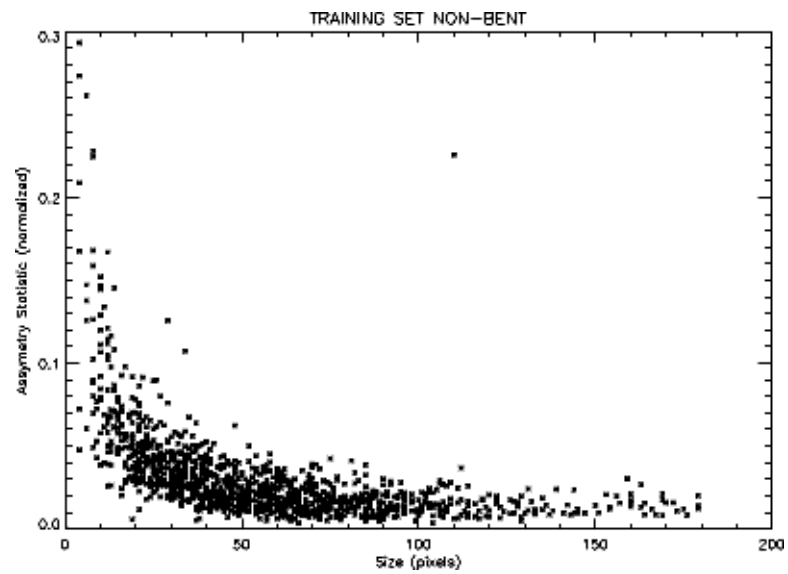
## TEST SET - Visual Classifications

52 Bent  
 165 Somewhat Bent  
 612 Non-Bent  
 227 ? (too noisy to tell)  
 3 Misc.

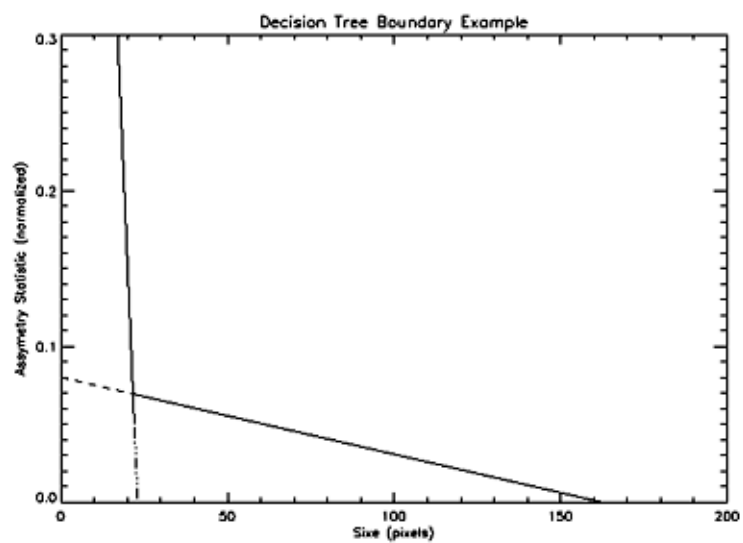
1059 Total

METHOD	BENTS FOUND (%)	FALSE POSITIVE (%)
OC1 decision tree & catalog features	71	75
OC1 decision tree & image statistics		
SNNs Rprop		



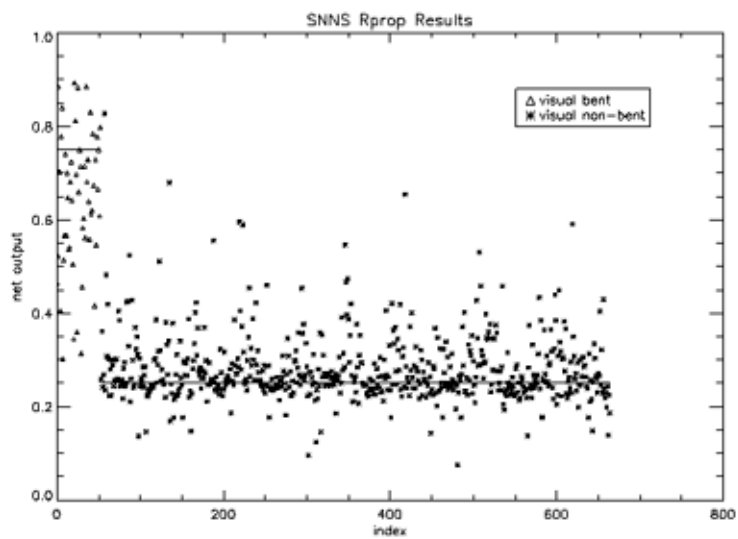






METHOD	BENTS FOUND (%)	FALSE POSITIVE (%)
OC1 decision tree & catalog features	71	75
OC1 decision tree & image statistics SNNs Rprop	88	54





METHOD	BENTS FOUND (%)	FALSE POSITIVE (%)
OC1 decision tree & catalog features	71	75
OC1 decision tree & image statistics	88	54
SNNS Rprop	85	25



Visual Classifications of 115 OC1 Selected  
Non-axisymmetric Sources:

27 Yes (NAT & WAT)  
21 Maybe  
15 Core-Jet  
52 No (Spiral & Other)

The future work is to automate these classifications.



# Noninvasive Recovery of Local Properties of Deforming Objects from Registered Range Data

UCRL-JC-135878ABS

Leonid V. Tsap(1), Dmitry B. Goldgof(2) and Sudeep Sarkar(2)

(1) Lawrence Livermore National Laboratory  
Center for Applied Scientific Computing

(2) Department of Computer Science and Engineering  
University of South Florida

Physically-based modeling involves the use of natural properties of nonrigid objects to determine motion parameters. What if some local material properties or geometry are unknown? Generalization and simplification attempts compromise the advantages gained from using such models. In this talk I present a noninvasive approach for iterative recovery of such initially unknown parameters. The method is based on the analysis of the differences between the actual behavior of the object (recovered from registered range and intensity image sequences) and its predicted behavior using a model. Large differences indicate that an object's properties are not captured properly by the model describing it. Feedback from the images during the motion allows for the refinement of the model by minimizing the error between the expected and true position of the object's points. Unknown parameters are recovered using an iterative descent search for the best nonlinear finite element model that approximates observed motion. As a result, we obtain a more precise description of the object including initially unknown local material properties or geometry.

Experimental results demonstrate the success of the proposed algorithm. The method was applied to man-made elastic materials and human skin to recover unknown elasticity, to complex 3-D objects to find details of their geometry, and to a hand motion analysis application. A very important application of the method is that it is capable of objective detection of the differences in elasticity between different materials as well as measuring burn scar elasticity and including it into the next, higher level of modeling such as our model of the hand. The latter is used for an assessment of the feasibility of applying combined computer vision and simulation techniques to the analysis of Repetitive Stress Injury (RSI). Experiments also show method's capabilities to detect the problems with location or size of specific parts of deformable objects. This work demonstrates the possibility of accurate quantitative analysis of nonrigid motion in range image sequences with objects consisting of multiple materials and 3-D volumes.

**Keywords:** physically-based vision, deformable models, nonrigid motion analysis, biomedical applications

*This research was supported in part by the Whitaker Foundation Biomedical Engineering Research Grant and in part by the National Science Foundation Grants IRI-9619240 and EIA-9729904.*

*This work was performed under the auspices of the U.S. Department of Energy by Lawrence Livermore National Laboratory under contract number W-7405-Eng-48.*



---

---

## Noninvasive recovery of local properties of deforming objects from registered range data

---

---

**Leonid V. Tsap <sup>(1)</sup>, Dmitry B. Goldgof <sup>(2)</sup> and Sudeep  
Sarkar <sup>(2)</sup>**

<sup>(1)</sup> Lawrence Livermore National Laboratory  
Center for Applied Scientific Computing

<sup>(2)</sup> Department of Computer Science and Engineering  
University of South Florida



November 11, 1999



## Motivation

---

---

- | **Physically-based modeling in vision and its requirements**
- | **The need to achieve precise motion analysis and object understanding given incomplete description of the object**
- | **The method uses:**
  - input: sequences of registered intensity and range images**
  - output: displacement fields, strain distributions and recovered geometry/ material properties**
- | **Applicational requirements in burn scar assessment, repetitive stress injury analysis and local shape estimation of deformable objects**

CASC

2



## Main steps of the method

---

- | Establish corresponding features from images
- | Detect areas with unknown properties:
  - create physically-based model
  - compute motion (by applying displacements)
  - analyze strain distribution
  - detect abnormal areas
- | Recover properties:
  - change affected property for such areas
  - compute model after the motion
  - estimate the error (differences with image data)
  - continue until minimum

CASC

3

## Experimental setup

---

- | The structured light system used in this research projects a sequence of horizontal stripes into the scene
- | The image is captured from a different viewpoint with a camera



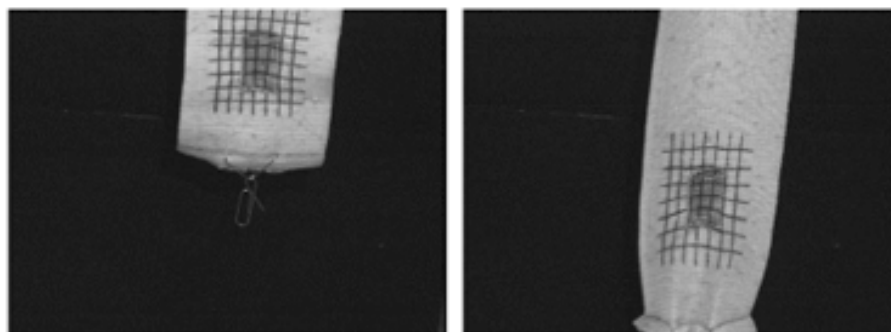
CASC

4



## Data acquired by the system

Intensity images of the elastic material before and after stretching. The grid painted on the bandage is used for tracking.

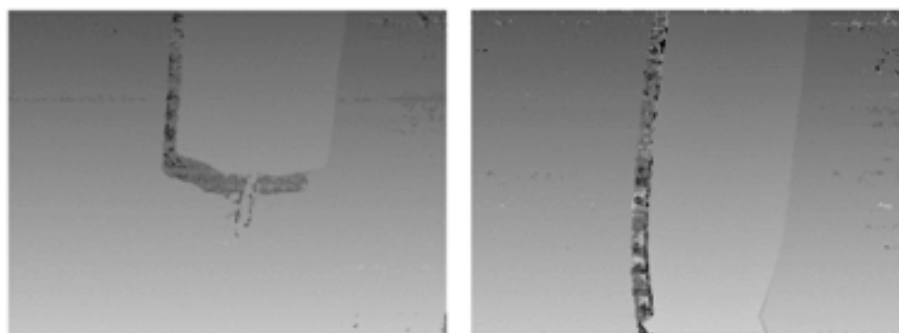


CASC

5

## Data acquired by the system

Range images of the elastic material before and after stretching



CASC

6



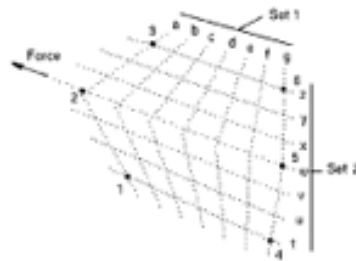
## Application of Active Contours

- | A snake is an energy minimizing spline, governed by internal and external forces. It is specified by a set of control points.
- | The energy function for a snake:  

$$E_{snake} = E_{internal} + E_{external}$$
- | Total snake energy definition for a greedy algorithm by Williams and Shah:  

$$\int (a(s) E_{cont} + b(s) E_{curv} + g(s) E_{img}(s)) ds$$

- | Snake placement

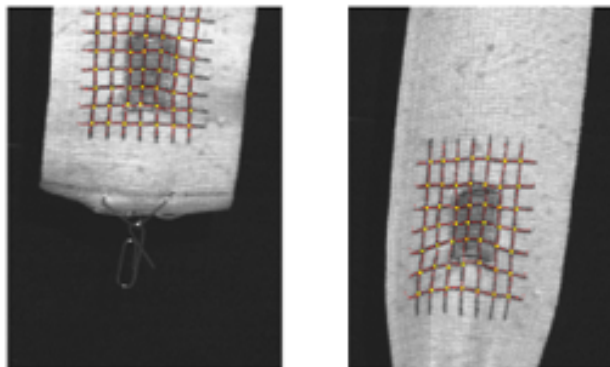


CASC

7

## Recovering material properties

Applying snakes to the intensity images (control points are red, snake intersections are yellow)



Detected intersections are used to build a finite element model

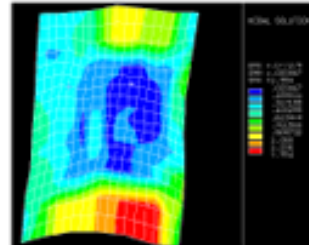
CASC

8

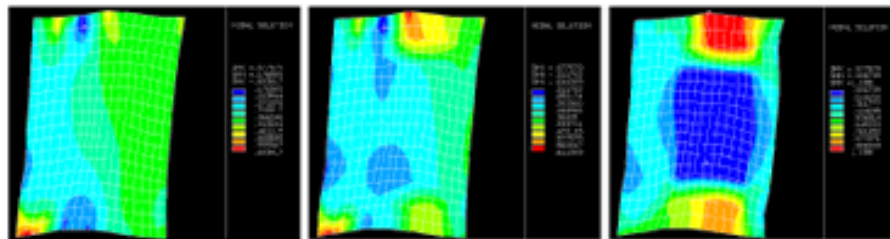


## Recovering material properties

Detection of abnormal area,  
pinpointed by the lowest  
strain (blue)



Iterative descent search for its  
properties (selected iterations)

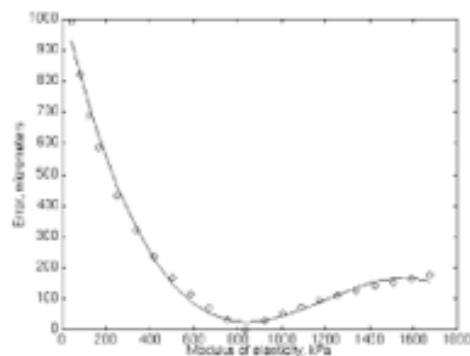


CASC

9

## Recovering material properties

Approximation of the error function (differences in  
displacements between the model and the object during  
the search). Minimum of the function corresponds to  
recovered elasticity of the abnormal area.



CASC

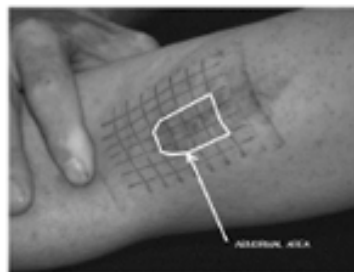
10



## Objective assessment of burn scars

- | Detection of scars from images
- | Evaluation of their elastic properties relative to the surrounding tissues
- | Ranking scars on the scale used by physicians
- | Providing an objective criteria for healing progress estimation and comparison of techniques

Similarly to elastic materials, distortions are evaluated when skin is pulled

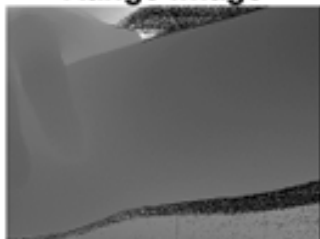


CASC

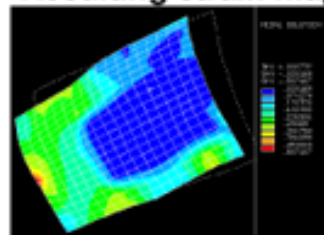
11

## Objective assessment of burn scars

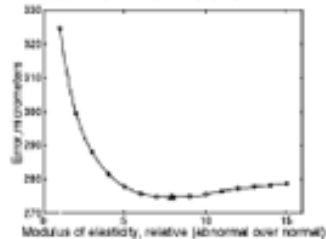
Range image



Resulting strain map

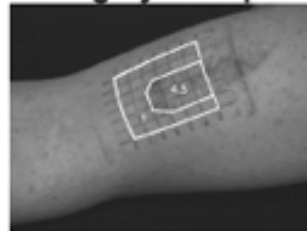


Error function



CASC

Scar rating by the specialist

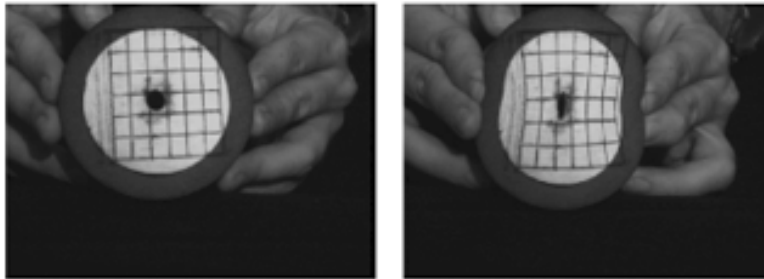


12



## Recovering missing geometry information

- | Often some info about local geometry is not available
- | Consider a cup holder sequence (range not shown)

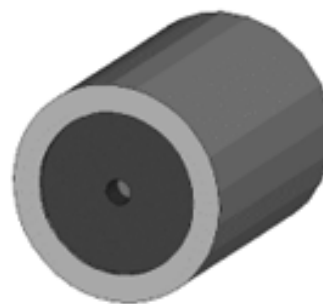
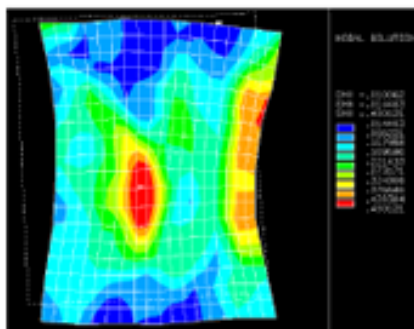


CASC

13

## Recovering missing geometry information

- | Area of interest is characterized by high strains (red)
- | Missing info (hole in the bottom of the cup holder) is recovered with a local model and integrated into a 3-D model



CASC

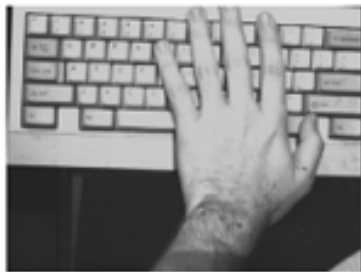
14



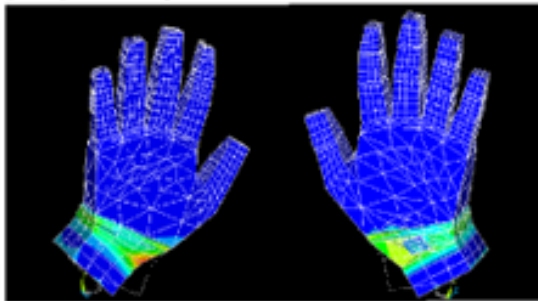
## Hand motion reconstruction and integration of recovered properties

---

- | Various positioning of the hands on the keyboard
- | Wide range of movements
- | Strains are projected onto the surface
- | Looking for excessive strains
- | Influence of burn scars on the hand motion (by including recovered properties)



CASC



15

## UCRL-VG-135878

---

Work performed under the auspices of the U. S. Department of Energy by Lawrence Livermore National Laboratory under Contract W-7405-Eng-48

CASC

16



## **Subsurface Gap Depth Detection by Infrared Imaging Using a Surface Heat Pulse**

*Charles Landram, Richard Martin, Nancy DelGrande and Philip Durbin*







---

Day 2

Opening

**Presentations**

---

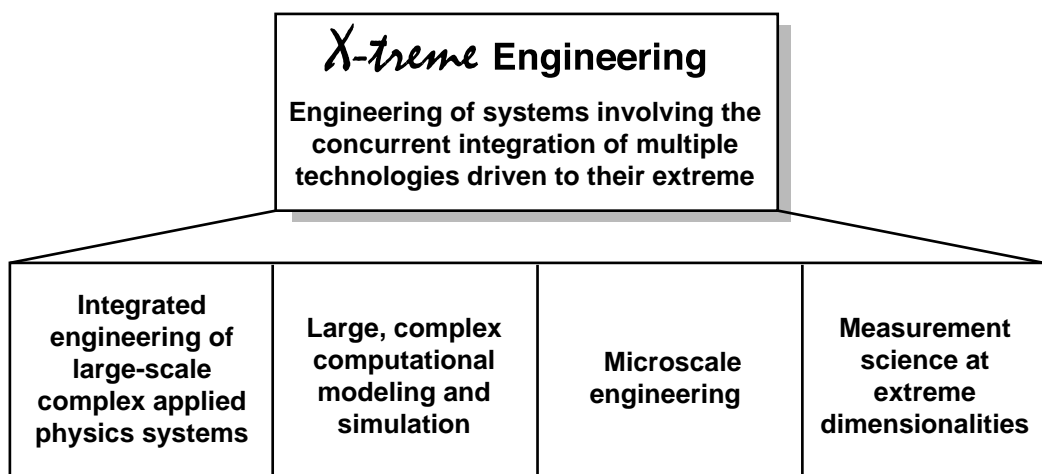


# Engineering Research and Development

**Spiros Dimolitsas**



## Engineering's Competencies





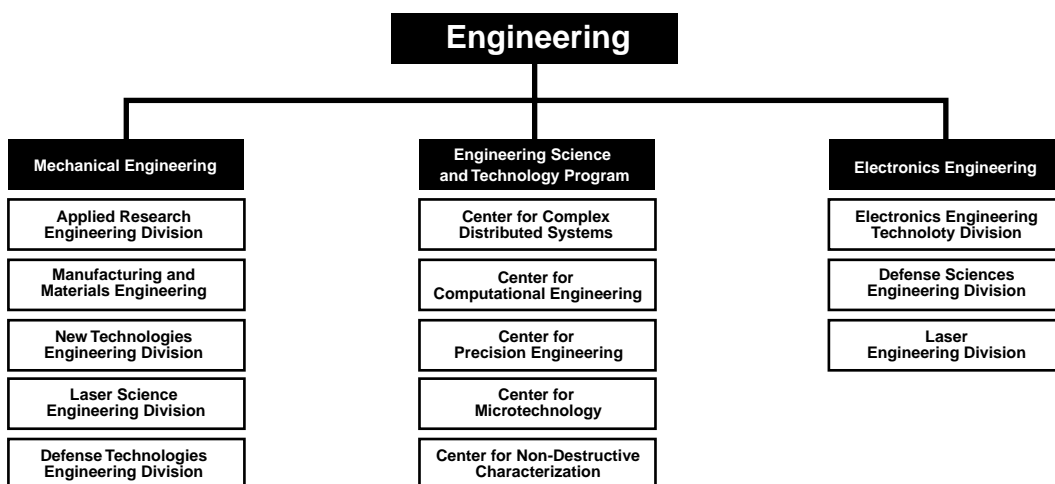
## Engineering's Core Technologies



- Systems Engineering
- Computational Engineering
- Precision Engineering
- Microsystems Technology
- Signal Acquisition and Characterization

**ENGINEERING**

## Engineering's Structure Reflects its Technologies



**ENGINEERING**



## Engineering Center Mission



- Engineering's Technology Centers are responsible for the vitality and growth of the core technology each of them represent:
  - They focus and guide Engineering's investments in enabling technologies and supporting capabilities
- Each Center is an Engineering R&D "program" whose output is technology
- Centers provide a forum for the discussion of issues related to the broader core technology area with which each of them is associated



## Each Center addresses a number of enabling technologies within their sphere of activity



### Computational Engineering

*Integrated modeling of assembled engineering systems*

### Precision Engineering

*Removal, addition, and measurement of material at near-atomic level*

### Micro-Technology

*Microscale fabrication of fluidic and photonic structures and systems*

### Non-Destructive Characterization

*Quantitative non-destructive characterization of components and systems at high resolution*

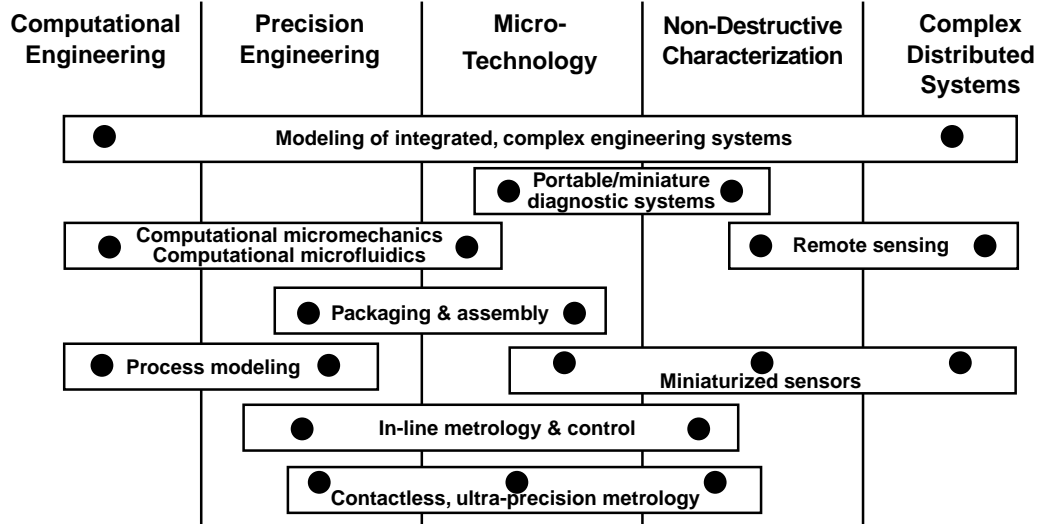
### Complex Distributed Systems

*Simulation, control and sensing of large-scale systems*





## Engineering's Core Technologies



**ENGINEERING** 



# A Retrospective on Computer Vision Research

*Avi Kak, Professor in the School of Electrical Engineering, Purdue University*

During the last thirty years, computer vision has gone through various paradigm shifts, but, alas, many of those paradigms have proved to be merely flashes in the pan. During this period, many slogans were advanced that supposedly represented fresh approaches to solving computer vision problems; slogans such as “active vision,” “purposive vision,” “reconstructive vision,” “knowledge-based vision,” “pure vision,” etc., were bandied about. But apart from giving us a new vocabulary for explaining why some of the problems may not be solvable in the near future, these slogans have not amounted to much.

On the other hand, the vision systems that actually work today in the leading laboratories of the world are based on sound engineering principles. These systems use time-proven signal and image processing algorithms for extracting low-level information from images, estimation-theoretic approaches for dealing with uncertainty, graph-theoretic approaches for representing the extracted information by relational data structures, tools of computer graphics for object modeling and rendering, etc. What’s interesting is that the success achieved in these systems is owing primarily to the workings of the underlying signal- and image-processing algorithms and to estimation-theoretic methods for coping with uncertainties. This is not to say that the high-level algorithmic control structures are dispensable, only that the systems tend to be more sensitive to the success of the signal-processing end. I will also try to show that when computer vision does not work, it is primarily due to the fact that we do not know how to devise appropriate signal and image processing strategies for low-level feature extraction and that sometimes intractable problems become solvable by incorporating a human in the loop for helping out with low-level feature extraction.

I will show results from some state-of-the-art experiments in model-based vision for object recognition, for indoor mobile-robot navigation, for content-based image retrieval, etc. The robustness achieved by combining model-based reasoning with uncertainty management in these systems has resulted in applications that are expected to benefit industrial automation and medical diagnostics of the future.



# Overview of Biopatheogen Detection Technology at LLNL

*Fred Milanovich*







---

# Processing for Complex Systems

---



# Overview of the Center for Complex Distributed Systems

*David McCallen*



---

## **Using signal processing techniques to improve finite element modeling**

---



**CASIS Signal and Image Sciences Workshop**

**November 11, 1999**

**Greg Burnett**

**Lawrence Livermore National Laboratory**

## **Finite Element Models (FEM)**

---



**Ubiquitous in engineering, used to model  
everything from beer cans to experimental  
airplanes to skyscrapers**

**Have come increasingly into use as computing  
power has become cheaper and more widely  
available**

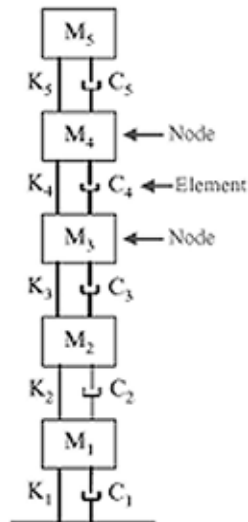
**They model continuous structures as a series of  
discrete elements that connect at nodes**

**Built using a standard set of elements and nodes  
plus a healthy portion of engineering judgement**

CASIS/NSA, 99 p. 2



## Typical low order FEM



- Fifth order system of a five story building
- Each element has a stiffness and shear value
- Each node connects elements and has its own properties
- Linear model characterized by  $M$ ,  $C$ ,  $K$  matrices, describes structure with a set of ODEs

CASIS No. 39 p. 3

## How are FEM traditionally tested and improved?



- Most of the testing involves the vibrational modes of the system under study
- For linear systems these are the eigenvectors and eigenfrequencies of  $M^{-1}K$
- The calculated modes of the FEM are compared to the experimentally measured modes of the structure
- Based upon the differences in the mode shapes and frequencies, the model can be changed in an effort to match the modal response

CASIS No. 39 p. 4



## More systematic methods of FEM improvement:

---



- **Gradient searches (Gauss-Newton)**
  - Good for signals with smooth minima
  - Can use with any cost function
- **Optimal Linear Filtering**
  - i.e. Kalman Filters. Good with noisy signals, more slowly converging
  - Essentially state estimators that can accept noisy measurements
- **Optimal matrix, sensitivity based parameter update, and many others**

CARBNv, 99 p. 5

## What is our research objective?

---



- **To improve Finite Element (FE) models systematically and effectively using state space-based parameter identification methods**

CARBNv, 99 p. 6





## Why is this useful?

---

- The FEM update procedure may be used for two purposes:
  1. To detect FEM mismatches (i.e. the model does not adequately describe the structure)
  2. To detect damage to a structure if an adequate model for the structure exists

CASIS No. 99 p. 7



## Why do we care?

---

- The possible applications of this technology are endless - FEM are extremely common in structural engineering
- Civil applications include the improved modeling of the response of bridges, buildings, and stadiums to events such as earthquakes or bombings
- Defense applications are many and varied

CASIS No. 99 p. 8





## What are the challenges of this project?

---

1. The models are usually large, consisting of a few hundred to a few thousand degrees of freedom
2. The number of measurements that we are able to make is usually much smaller than the DOF, on the order of 10-100
3. The measurements are usually noisy

CASIS No. 99 p. 9



## What are the challenges of this project?

---

1. The models are usually large, consisting of a few hundred to a few thousand degrees of freedom  
This can result in high computational expense
2. The number of measurements that we are able to make is usually much smaller than the DOF, on the order of 10-100
3. The measurements are usually noisy

CASIS No. 99 p. 10





## What are the challenges of this project?

---

1. **The models are usually large, consisting of a few hundred to a few thousand degrees of freedom**

This can result in high computational expense

2. **The number of measurements that we are able to make is usually much smaller than the DOF, on the order of 10-100**

This can cause observability and controllability problems - it is difficult to tell where the damage is precisely

3. **The measurements are usually noisy**

CASIS No. 99 p. 11



## What are the challenges of this project?

---

1. **The models are usually large, consisting of a few hundred to a few thousand degrees of freedom**

This can result in high computational expense

2. **The number of measurements that we are able to make is usually much smaller than the DOF, on the order of 10-100**

This can cause observability and controllability problems - it is difficult to tell where the damage is precisely

3. **The measurements are usually noisy**

This can impede the ability of the algorithms to converge on the correct answer

CASIS No. 99 p. 12





## How do we overcome the difficulties?

---

1. The models are usually large, consisting of a few hundred to a few thousand degrees of freedom
2. The number of measurements that we are currently able to make is usually much smaller, on the order of 10-100
3. The measurements are usually noisy

CASIS Nov. 99 p.13



## How do we overcome the difficulties?

---

1. The models are usually large, consisting of a few hundred to a few thousand degrees of freedom
  - Reduce the model order through intelligent (human based) condensation
  - Use recursive algorithms that process data sequentially, reducing the amount of memory required
2. The number of measurements that we are currently able to make is usually much smaller, on the order of 10-100
3. The measurements are usually noisy

CASIS Nov. 99 p.14





## How do we overcome the difficulties?

### 1. The models are usually large, consisting of a few hundred to a few thousand degrees of freedom

Reduce the model order through intelligent (human based) condensation

Use recursive algorithms that process data sequentially, reducing the amount of memory required

### 2. The number of measurements that we are currently able to make is usually much smaller, on the order of 10-100

Estimate the unmeasured data with the model and place sensors at positions where there is the most uncertainty

Interpolate the measurements intelligently

### 3. The measurements are usually noisy

CASIS No. 99 p. 13



## How do we overcome the difficulties?

### 1. The models are usually large, consisting of a few hundred to a few thousand degrees of freedom

Reduce the model order through intelligent (human based) condensation

Use recursive algorithms that process data sequentially, reducing the amount of memory required

### 2. The number of measurements that we are currently able to make is usually much smaller, on the order of 10-100

Estimate the unmeasured data with the model and place sensors at positions where there is the most uncertainty

Interpolate the measurements intelligently

### 3. The measurements are usually noisy

Use a statistically based estimation algorithm

CASIS No. 99 p. 16



# Damage Detection

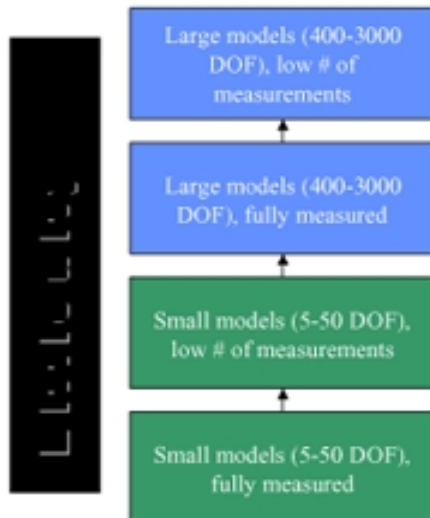
- **Four levels of damage detection**
  - Level 1: Is the system damaged?
  - Level 2: Where is the damage located?
  - Level 3: What type of damage is present?
  - Level 4: What is the extent of the damage?
- Historically only levels 1-2 addressed, large amounts of damage needed to localize
- Usually done by comparing mode shapes and frequencies, but some signal processing techniques such as LPC modeling used

# System Identification

- System identification involves level 3-4 damage detection — determining the amount and type of damage
- Output from accelerometers on actual structures is used to update FEM with values of K that are more appropriate to the structure
- Very little can be done presently to update models systematically — especially with low observability. It is a mix of art and science.



## Steps of progress for system ID



¥We are approaching the problem systematically, using smaller models with simulated data to wring out the algorithms

¥After the algorithms have proven themselves in simulations, we will process real data from physical structures

CASIS Nov. 99 p. 19

## We have made extensive use of a damped Gauss-Newton search algorithm in our initial studies



It uses a Gaussian approximation to the Newton method to search for a minimum of the least-squares search criterion

### Its advantages are:

- Quickly converges on relatively linear minima with small residuals
- Guaranteed to converge on a minima
- Widely used and available (Matlab's PEM is similar)

### Its disadvantages are:

- May not converge to a global minima
- May not converge if residuals large, system too nonlinear
- Noise can prevent convergence

CASIS Nov. 99 p. 20



## So what have we learned?



**Parameter identification is possible for both fully measured and partially measured systems, but it is critical to select the correct measurement nodes**

**For large DOF systems, Gauss-Newton search may not be the best algorithm**

—High DOF systems span large dimensional space, many locals minimums possible

CARS No. 99 p. 21

## Example illustrating the effect of sensor placement on identification accuracy (10 DOF, 5 rotational, 5 translational, sim data, no noise)



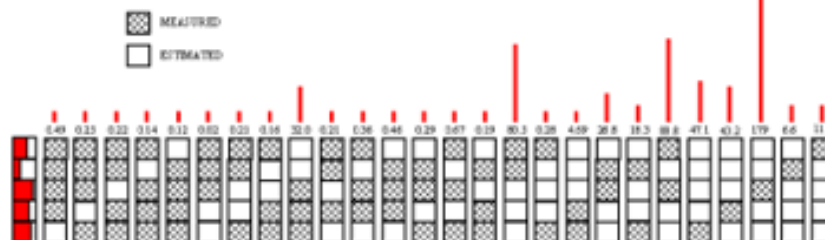
We were able to detect large (up to 2/3 of the original value) perturbations

With 5 measurements accurate to better than 0.01%

With 4 measurements accurate to better than 1%

With 3 measurements accurate from less than 1% (9 cases) to 32% (1 case), depending on states measured

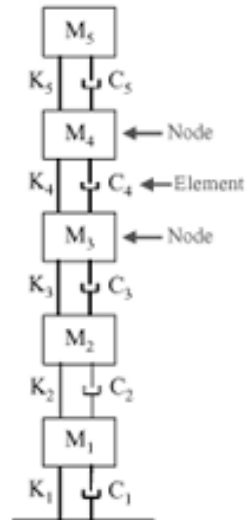
With 2 or 1 measurements, the mean error could increase to 80% or more



CARS No. 99 p. 22



## Typical low order FEM



**Fifth order system of a five story building**

**Each element has a stiffness and shear value**

**Each node connects elements and has its own properties**

**Linear model characterized by M, C, K matrices, describes structure with a set of ODEs**

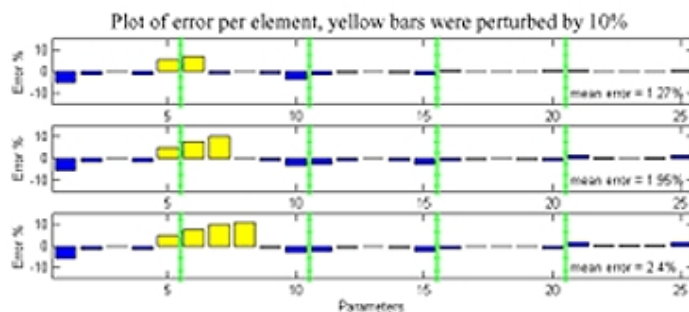
CASIS Nov. 99 p. 23

## 50 DOF results (only 5 translational states measured)



**Good news:** Can find and partially correct element errors near measurement points

**Bad News:** Elements far from measurement points basically unaffected



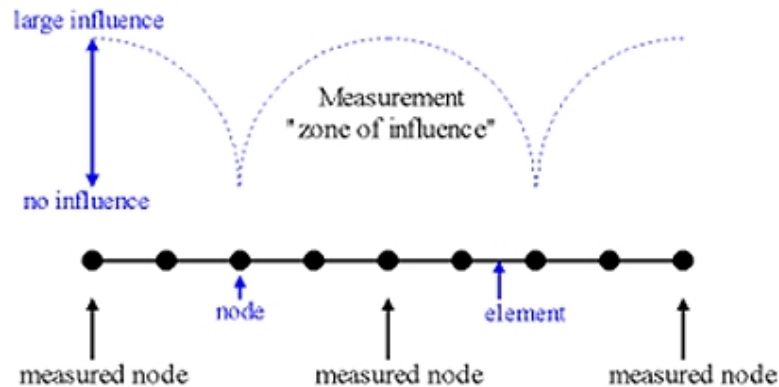
CASIS Nov. 99 p. 24



**For systems that are not measured at every node, damage detection and updates for nonmeasured nodes are difficult.**



**Instead of a direct connection, there exists a zone of influence around the measured node**



CMB/NW, 99 p. 25

## **Conclusions on simulated system ID**



**Accurate parameter identification (and thus levels 3-4 of damage detection) is only possible near the measurement points**

**How close it is necessary to be to the measurement points has not been determined, right now elements within about 1-2 elements from the measured node can be accurately identified**

**This distance is variable depending on the algorithm used**

CMB/NW, 99 p. 26



**DAMAGE DETECTION  
FOR  
ENHANCED EVALUATION OF STRUCTURES**

*G. Clark, G. Burnett, McCallen*

CASIS Workshop

Livermore, CA  
November 11-12, 1999

LAWRENCE LIVERMORE NATIONAL LABORATORY

## ORGANIZATION

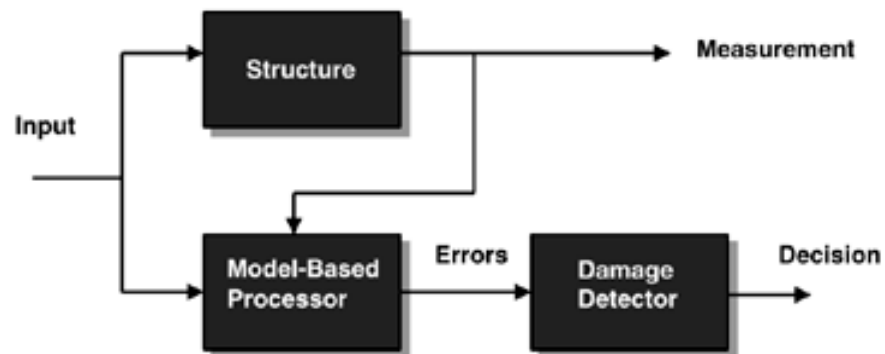
---

- INTRODUCTION
- MODELS
- DAMAGE DETECTION
- PROCESSING RESULTS
- SUMMARY

LAWRENCE LIVERMORE NATIONAL LABORATORY



**Damage detection is achieved using a model-based processor**



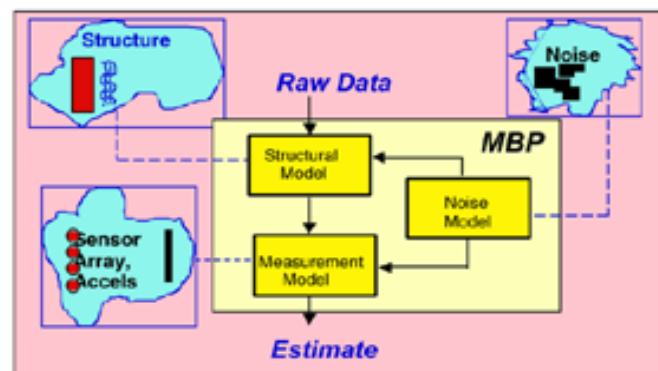
LAWRENCE LIVERMORE NATIONAL LABORATORY

**MBP design for vibrating structures requires that:**

we seek techniques to incorporate:

- \* structural models
- \* sensor (measurement) models
- \* noise models (ambient and measurement)

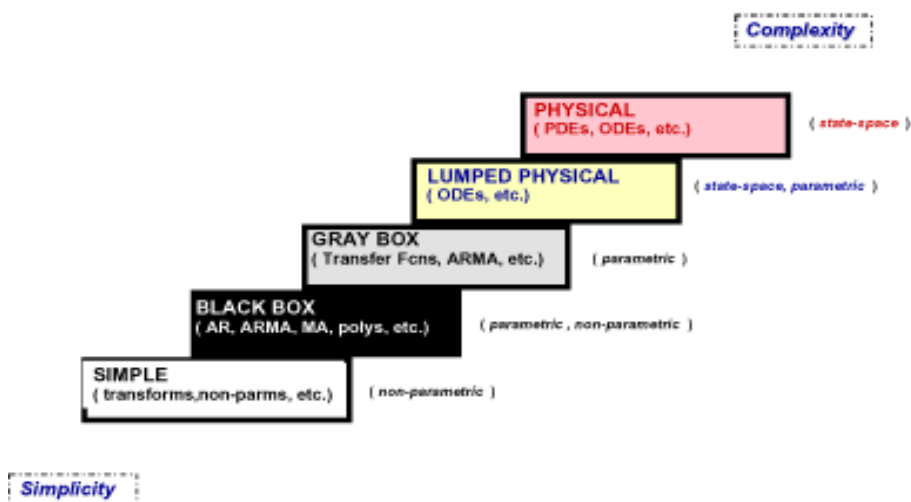
into the processor



LAWRENCE LIVERMORE NATIONAL LABORATORY

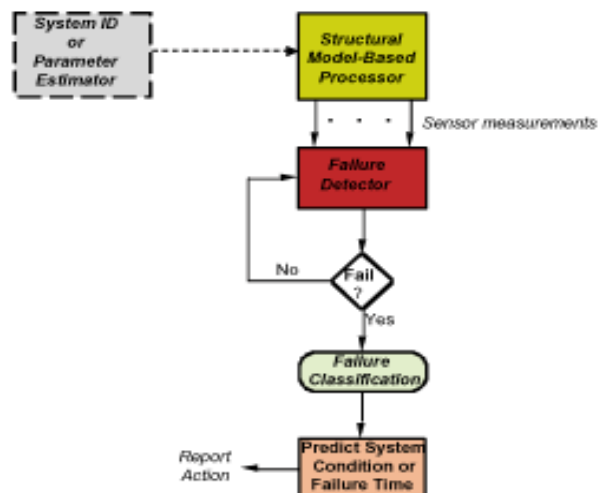


# The choice of MODEL determines the processing algorithm:



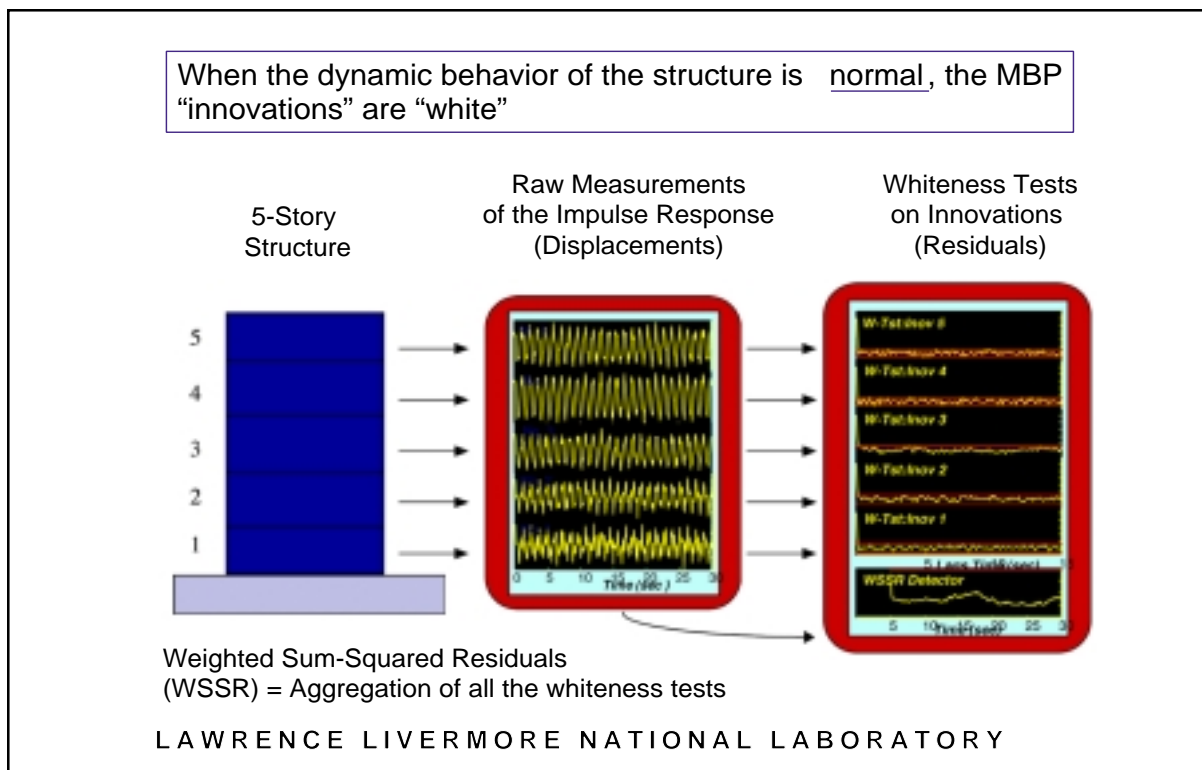
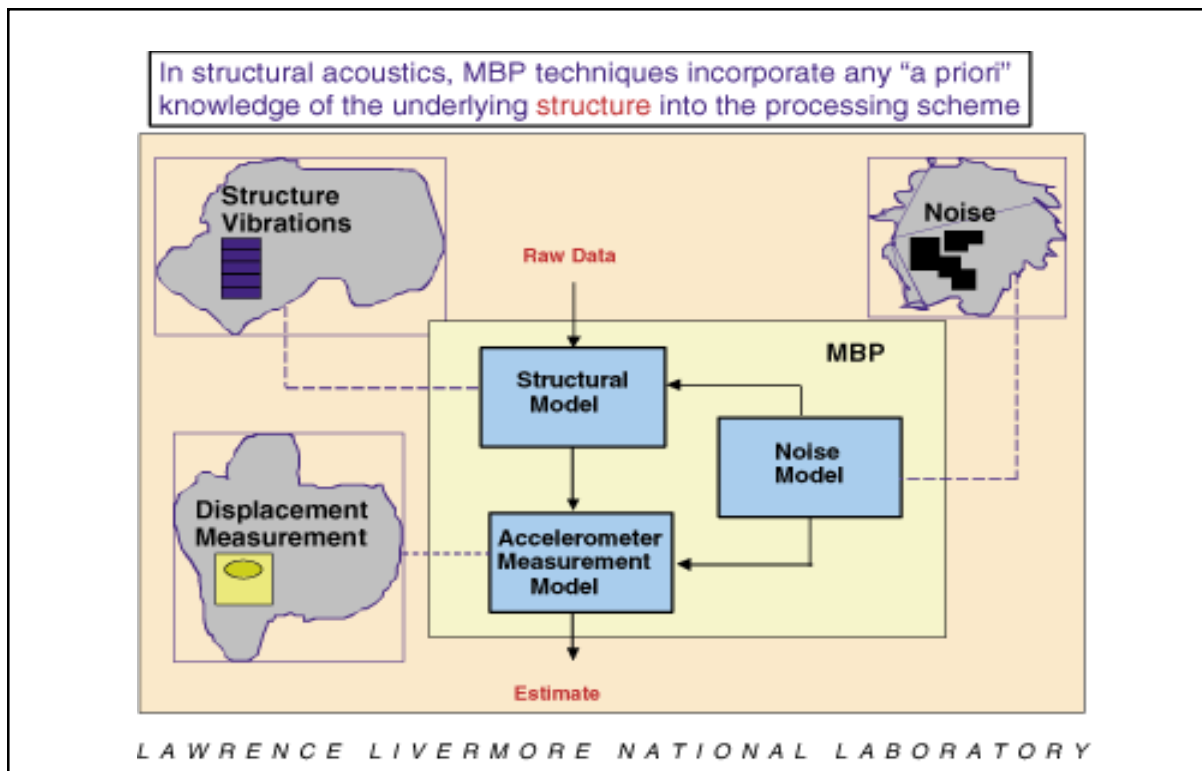
LAWRENCE LIVERMORE NATIONAL LABORATORY

# The detection, classification and prediction of structural failures is possible with the model-based approach



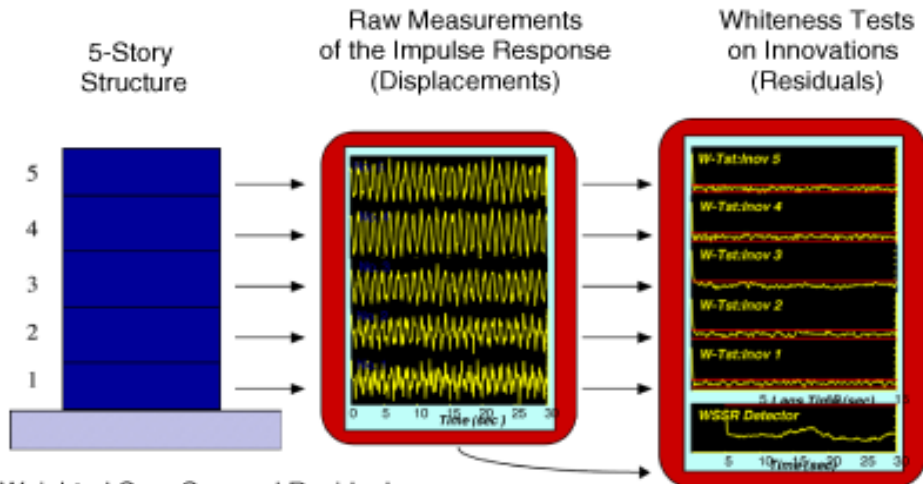
LAWRENCE LIVERMORE NATIONAL LABORATORY







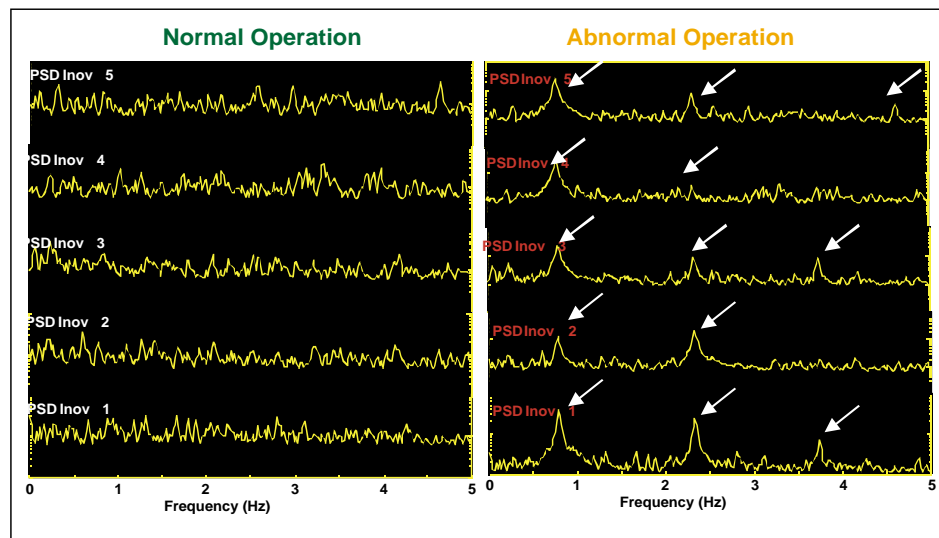
When the dynamic behavior of the structure is normal, the MBP "innovations" are "white"



Weighted Sum-Squared Residuals (WSSR) = Aggregation of all the whiteness tests

LAWRENCE LIVERMORE NATIONAL LABORATORY

Comparing PSDs of the innovations (residuals) also shows the damage resonances



LAWRENCE LIVERMORE NATIONAL LABORATORY



## Parabolic Reflector for Radar Mine Detection

S. Azevedo, T. Rosenbury, G. Dallum, M-L Liu, M. Vigars,  
Lawrence Livermore National Laboratory(LLNL)

Jamie Gough  
U.C. Davis/LLNL



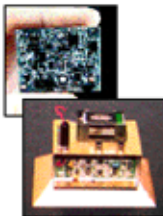
Prof. Carey Rappaport, ECE Dept  
Northeastern University



CASIS Workshop, LLNL  
November 11-12, 1999

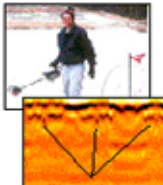
### LLNL has been researching radar mine detection techniques using new technologies

MIR Sensors



- Micropower Impulse Radar (MIR)
  - Extremely small, low power, inexpensive radar
  - Developed for LLNL's Laser Fusion Program
  - Now being applied to numerous government and commercial problems
  - Can be assembled into arrays

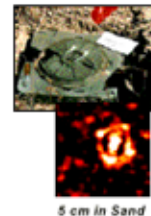
Hand-held Sensors



Waterfall Display

- Single-element and SAR configurations completed
- Has been tested at mine fields at LLNL, Nevada Test Site, Fort Belvoir, Fort A.P. Hill with success
- More data required for PD & FAR measures, but results are encouraging

Anti-tank M19



5 cm in Sand



3 Mines Localized in Clutter

This is the smallest known radar, hence it can be assembled into imaging arrays



CASIS 11.12.99  
SAETR 2



## Outline

---

- Mine detection problem -- forward-looking, light-weight AP mine detector
- Ground-clutter minimization -- plane wave vs. point source impulse excitation
- Offset paraboloid reflector
- FDTD modeling
- Multistatic receiver array for time-domain focusing
- Antenna fabrication
- Experimental results
- Next steps



CASIS 11.12.09  
SAETR 3

## A strong specular ground reflection hampers the effectiveness of radar mine detection

---

- Specular reflection from the ground depends only on incidence angle, frequency content, and wave polarization, not position or source.
- For flat ground with non-normal incidence, reflection scatters in forward direction.
- Plane wave excitation generates scattered signals that are easier to distinguish from rough surface clutter.
- Parabolas convert circular waves to plane waves.
- In the nearfield, a paraboloidal reflector generates a (non-uniform) plane wave across its aperture.
- An offset paraboloidal section must be used to avoid blockage and scattering by transmitting feed.
- Trade-offs between reflector size and grazing incidence angle -- and height above ground versus forward look -- must be made.

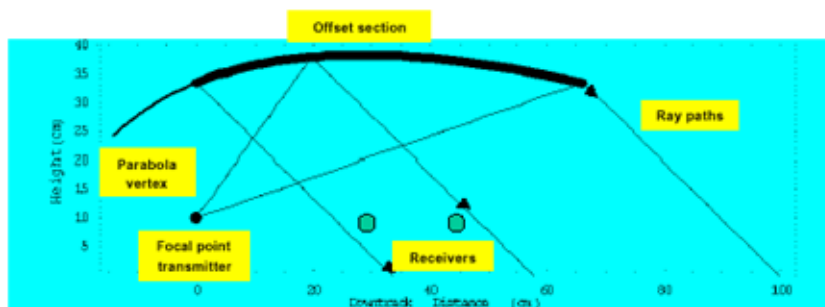


CASIS 11.12.09  
SAETR 4

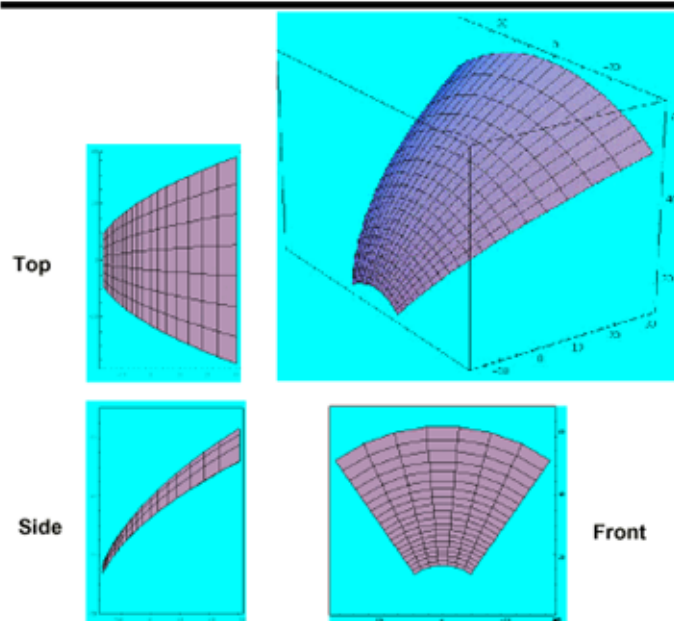


## A novel Forward-Looking "Handheld" AP Mine Detector with Plane Incidence Wave Excitation

### Offset Paraboloid Ray Paths



### 3D views of offset paraboloid

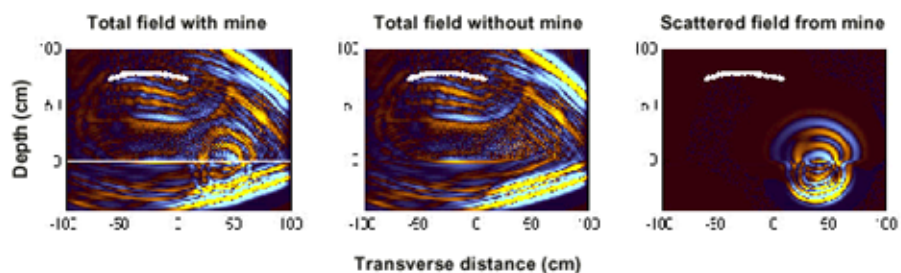


CASIS 11.12.99  
SAETRE 6



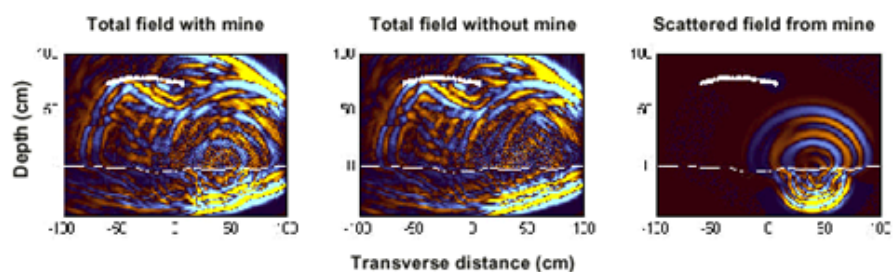
## 2-D FDTD Simulation of Parabolic Reflector Generated V-Pol. Planar Incident Wave

### Non-Metallic Mine 10 cm Below Flat Surface



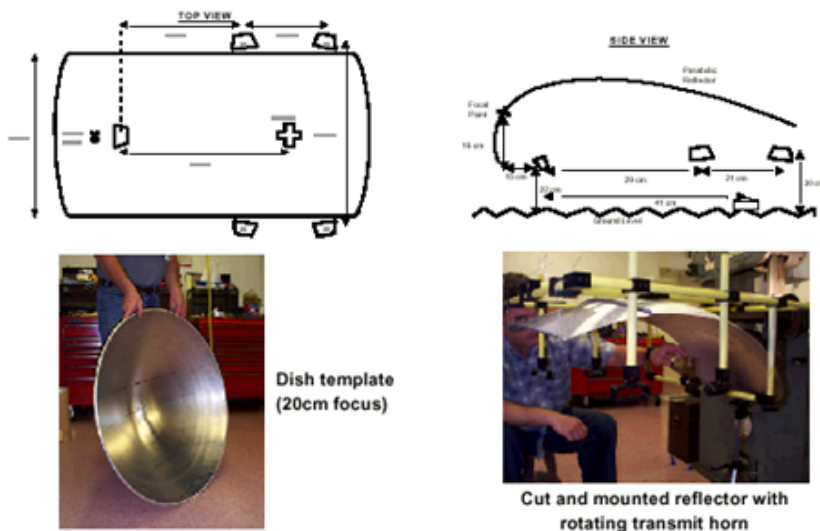
## 2-D FDTD Simulation of Parabolic Reflector Generated V-Pol. Planar Incident Wave

### Non-Metallic Mine 10 cm Below Rough Surface





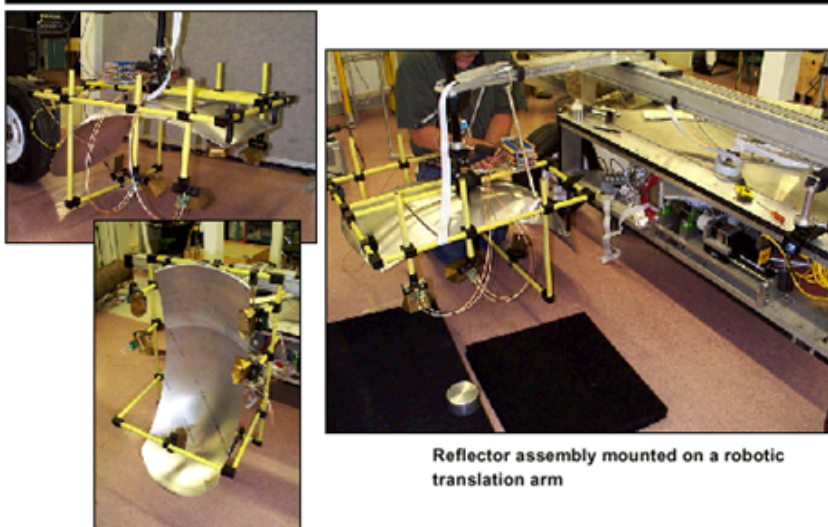
### The parabolic reflector system was built and tested as a prototype detection device



Lawrence Livermore  
National Laboratory

CASIS 11.12.99  
SAETR 9

### Four micropower radar receivers, in array configuration, were mounted on the reflector

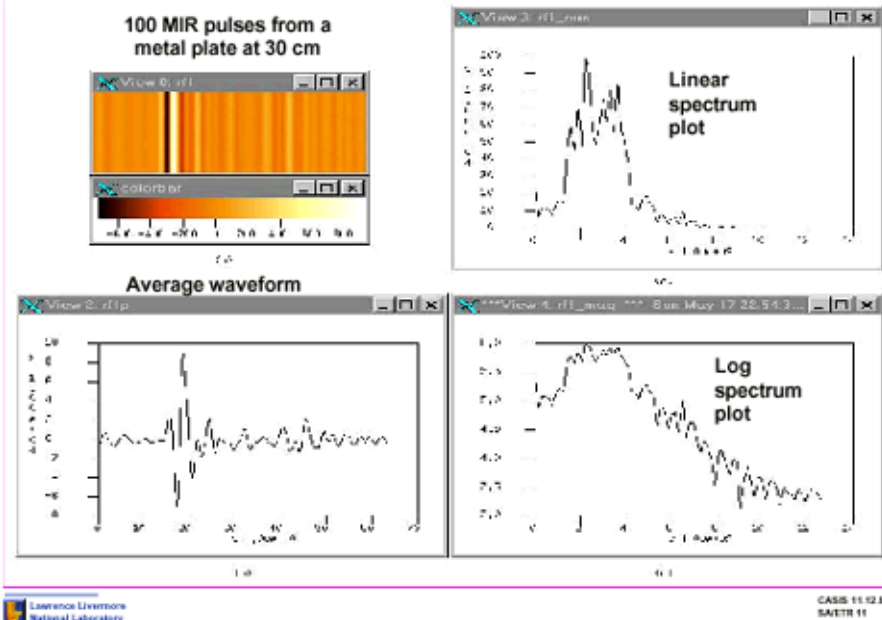


Lawrence Livermore  
National Laboratory

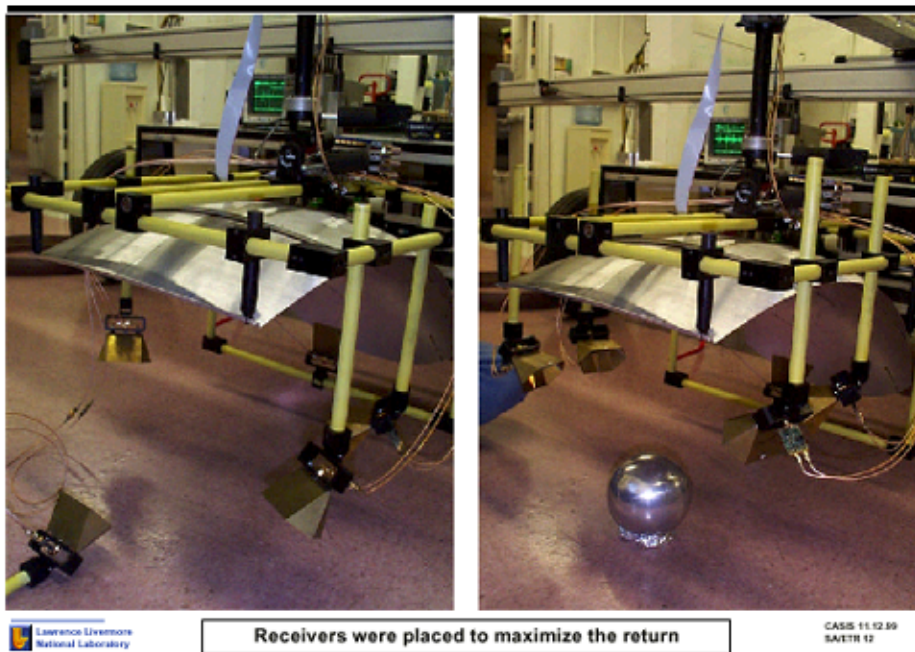
CASIS 11.12.99  
SAETR 9



## The MIR pulse is stable and wideband



## Calibration showed near-constant illumination in the field of view





### Sphere and mock-mine objects were tested in three soil types – sand, gravel, top-soil (dirt)



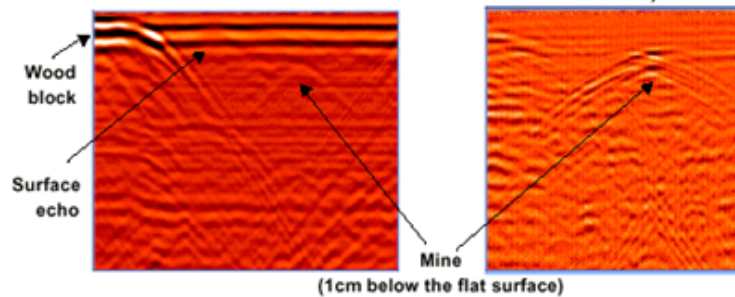
Lawrence Livermore  
National Laboratory

CASIS 11.12.99  
SAETR 13

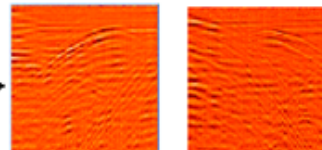
### The first-surface echo is greatly reduced with the reflector vs a monostatic radar

Monostatic radar aimed down at the soil from 10cm

Parabolic array radar aimed 45° to the soil (combined left/right receivers)



Raw signatures from left and right receivers



Lawrence Livermore  
National Laboratory

CASIS 11.12.99  
SAETR 56

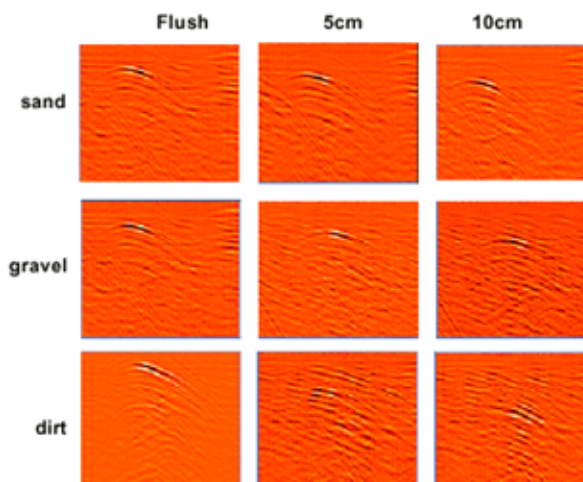


## A metal sphere was used for calibration



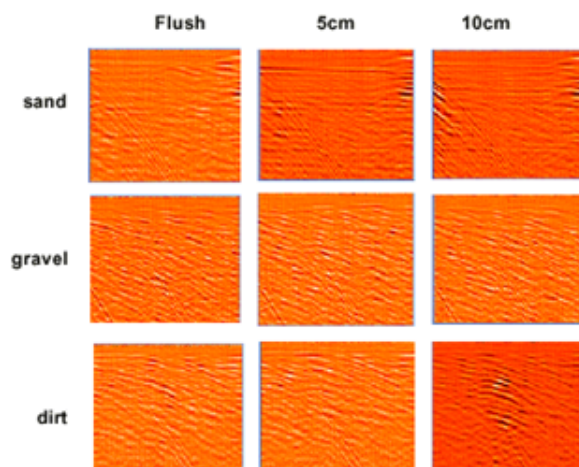
Inert M-14  
Metal sphere

Radar 1 returns for the sphere buried:



## Results in 2D for the inert mine will need improvement

Radar 1 returns for the mine buried:



Possible areas of improvement:

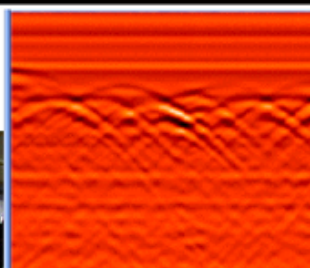
- Power
- Receiver placement
- Polarization
- Contrast enhance
- Reconstruct image



# Rough-surface tests in the sand showed the most possibility for S/N improvement



Lawrence Livermore  
National Laboratory

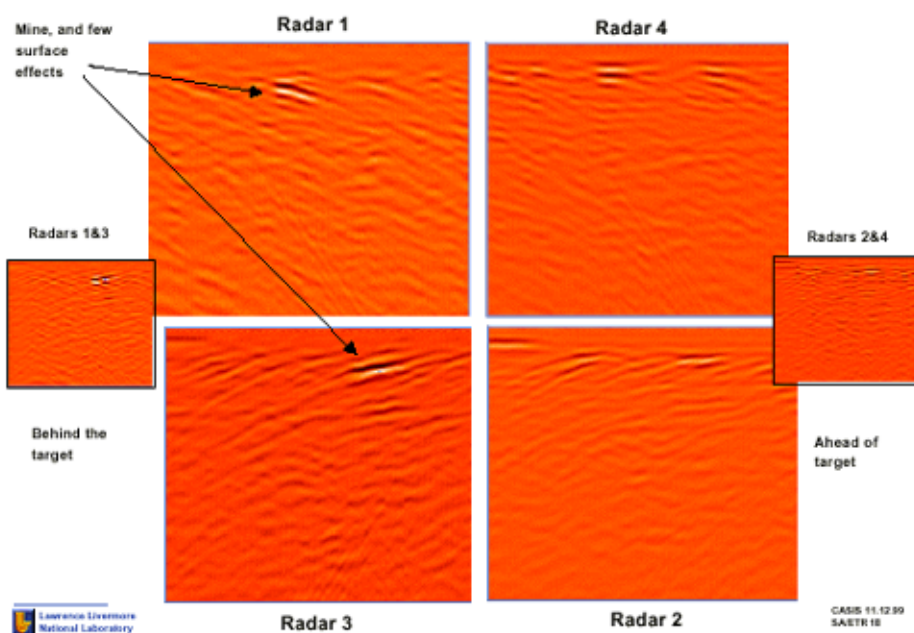


Monostatic radar results  
above

Inert mine buried 1cm  
below surface

CASIS 11.12.99  
SAETR 17

# Rough-surface tests in the sand showed the most possibility for S/N improvement (cont)



Lawrence Livermore  
National Laboratory

CASIS 11.12.99  
SAETR 18



### **Preliminary results show promise for the NU-designed parabolic reflector array antenna**

---

- Reflector provides good gain over large area
- Wavefront is measured to be planar in that area
- First surface echo is reduced greatly (over monostatic results)
- Placement and orientation of receiver antennas has an effect, but more research is needed
- Combining the delayed receiver waveforms improves the signal-to-clutter ratio
- Metal sphere (for calibration) can be detected in all soils at all depths
- Rough surface results show improved detection with reflector
- Image reconstruction algorithms need to be developed for the reflector geometry and to improve the detection



# Broad Band Acoustic Ranging and Velocimetry

*David Chambers, Karl Fisher, Brian Guidry, David Erskine*

Acoustic ranging has a long history of development beginning with sonar in World War II and continuing with the advent of medical ultrasound. Acoustic velocimetry (velocity measurement) is not as common but has been used to measure blood flow in the body. Simultaneous measurement of range and velocity is used in some medical ultrasound imaging but the resolution is coarser than with conventional medical imaging (ranging only). In the conventional approach to acoustic ranging and velocimetry, short broad bandwidth pulses provide better range resolution while long narrow bandwidth pulses provide better velocity estimates.

In this work we apply a new optical technique developed by Dave Erskine for white light ranging and velocimetry to the acoustic case. We generalize the technique to coherent broad band pulses and test it with spheres pulled through a specially designed water tank. We compare the sensitivity of the broad band technique with the conventional narrow band approach for velocity measurements. These experiments show that both range and velocity can be accurately measured with broad band pulses without degrading the range resolution. The technique can be easily applied to conventional sonar or radar, and medical imaging.

---

## Broad Band Acoustic Ranging and Velocimetry

David Chambers, Karl Fisher, Brian Guidry,  
David Erskine

CASIS Workshop  
November 12, 1999



**Conventional ranging and velocimetry trade  
accuracy of range estimates with accuracy of  
velocity estimates**

---

- | Accurate range estimates favor broad band (short) pulses
- | Accurate velocity (Doppler shift) estimates favor narrow band (long) pulses
- | Current systems (e.g. Doppler medical ultrasound) trade resolution (range estimation) for Doppler sensitivity
- | Sonar systems typically estimate velocity by tracking range over time.

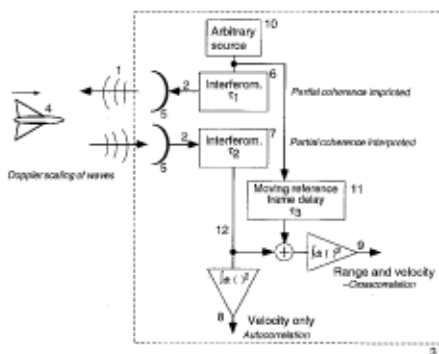
**Erskine's white light velocimetry method allows  
accurate velocity measurements with broad band  
pulses**

---

- | Use an interferometer to precondition broad band incoherent (white) radiation to illuminate object
- | Pass reflected radiation through a matching interferometer to detect phase difference induced by object motion
- | Measure time delay for range information
- | Preconditioning allows accurate measurement of phase shifts in broad band pulses with good range resolution



### Schematic of white light velocimetry system



### Goals of LDRD feasibility study

- ▮ Determine accuracy of range and velocity estimation using Erskine's original concept for an acoustic system
- ▮ Apply concept to coherent broad band acoustic pulses
- ▮ Determine accuracy limits for joint range and velocity estimation for this approach
- ▮ Assess the generality of this concept through better understanding of the role of preconditioning



### Acoustic implementation of white light velocimetry concept

Transmit basic pulse  $p(t)$  and time delayed version:  $p(t) + p(t-t_1)$

Reflected pulse is  $p_r(t) = A \left[ p(g(t-t_1)) + p(g(t-h(t-t_1))) \right]$ ,

$$t_1 = \frac{2r}{c}, \quad h = l + \frac{2U}{c}, \quad g = \frac{l+U/c}{l-U/c}$$

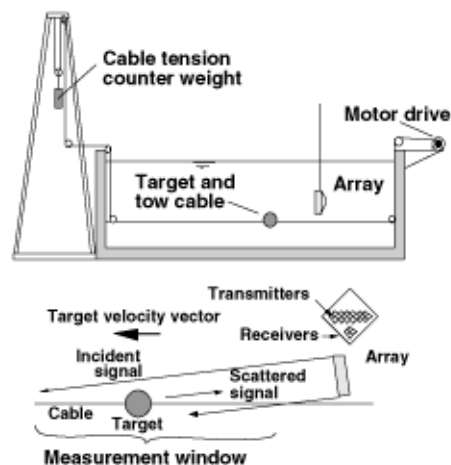
Add delayed version of reflected pulse to itself:  $p_r(t) + p_r(t-t_2)$

Calculate intensity as function of second delay

$$I(t_2) = \int \left[ p_r(t) + p_r(t-t_2) \right]^2 dt$$

Intensity peaks when  $t_2 = h t_1$ , then  $U = \frac{1}{2} c (t_2 - t_1) / t_1$

### Experimental setup



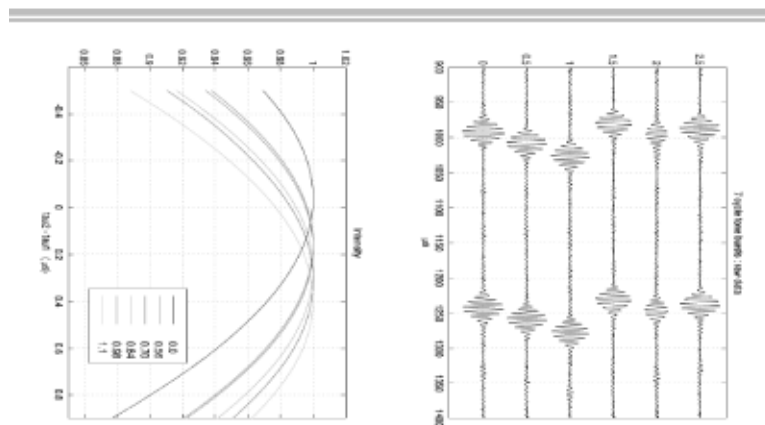
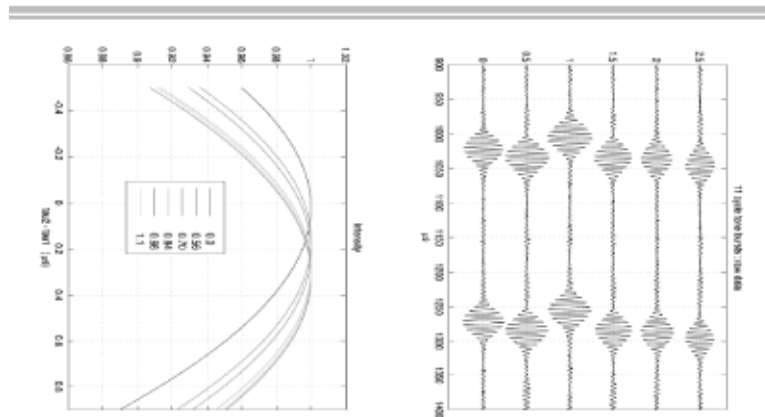


### Experimental cases

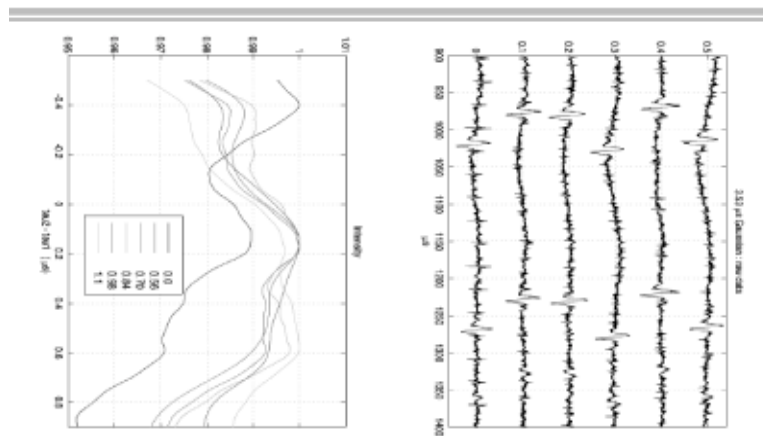
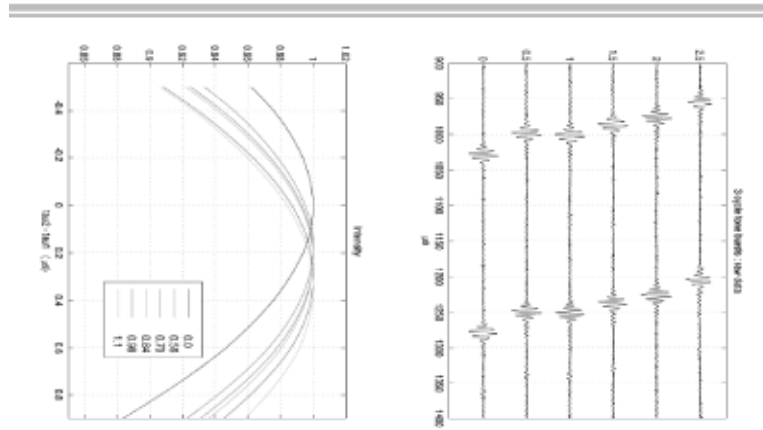
- ▮ Tested three Gaussian pulses with half widths 0.706, 1.18, and 3.53 ms, and three 150 kHz tone bursts of length 3, 7, and 11 cycles
- ▮ Towed air filled sphere at six speeds: 0.0, 0.56, 0.70, 0.84, 0.98, and 1.1 m/s
- ▮ Pulses generated by arbitrary waveform generator with  $\Delta t = 0.1$  ms
- ▮ Initial delay  $t_i$  is 250 ms
- ▮ From condition  $|t_2 - t_1| > \Delta t$  we have  $U > \frac{c \Delta t}{2 t_i} = 30 \text{ cm/s}$



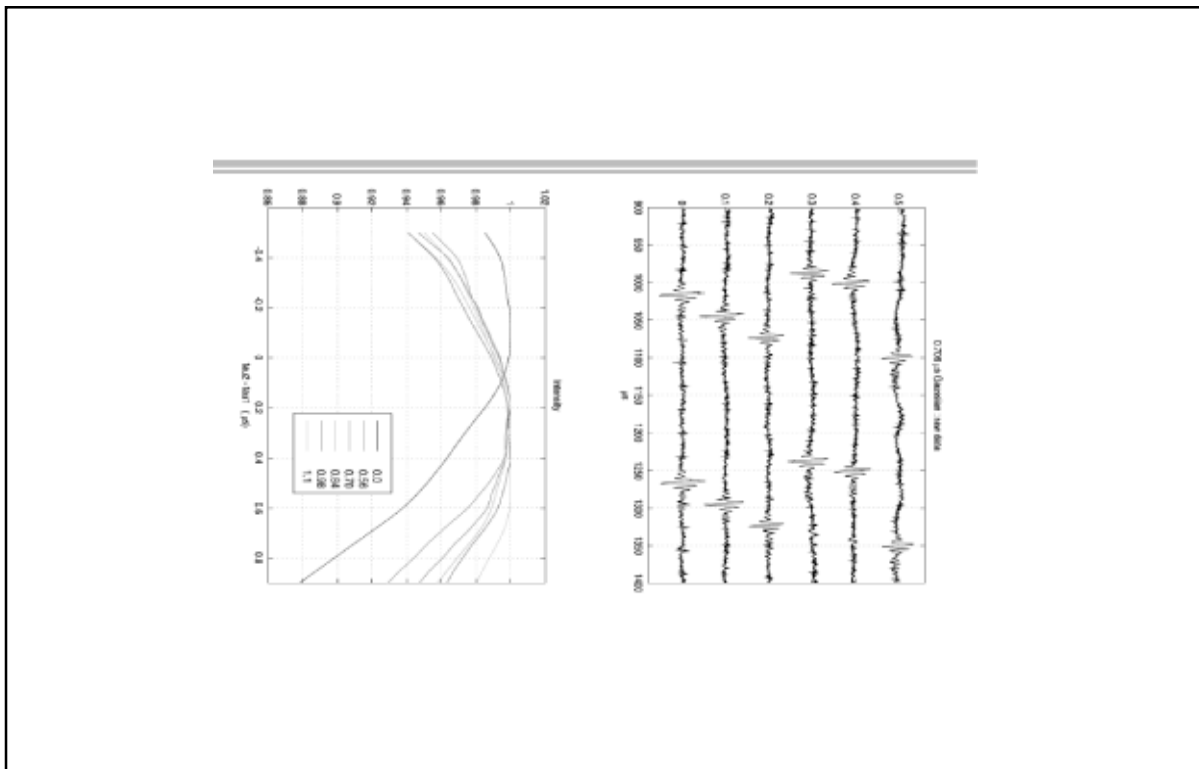
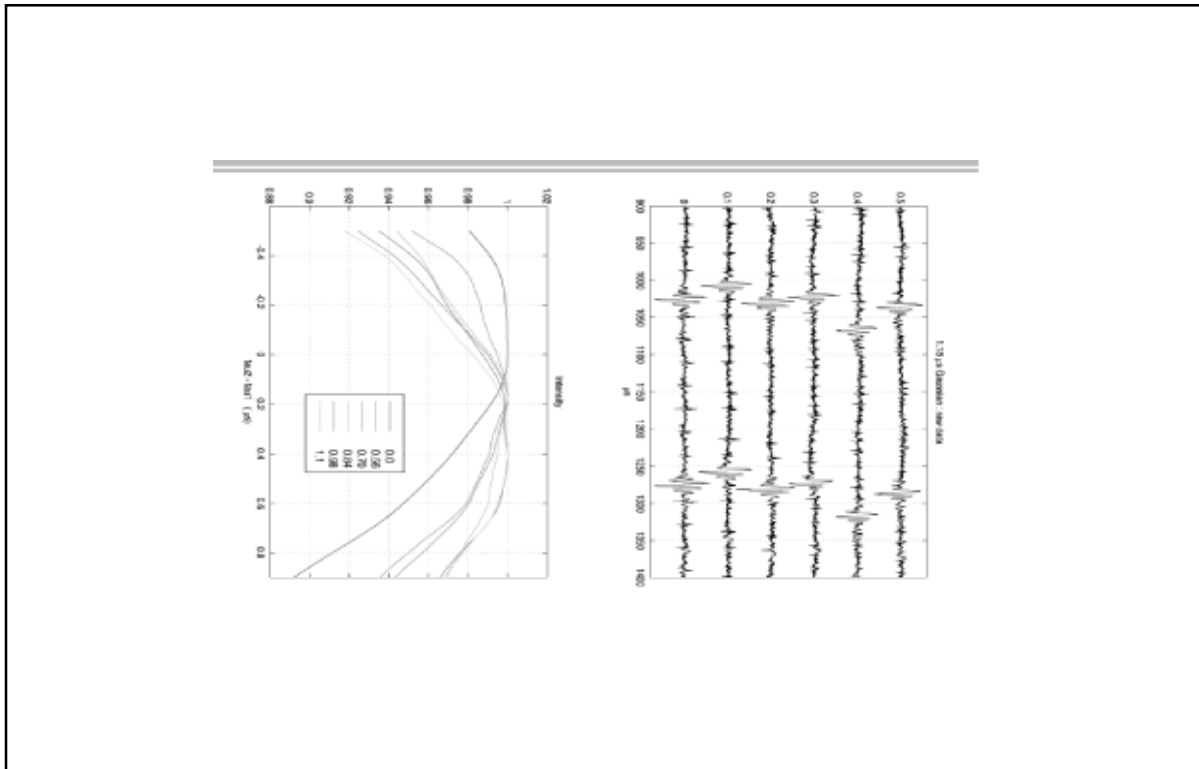






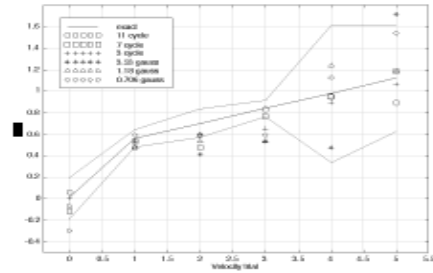




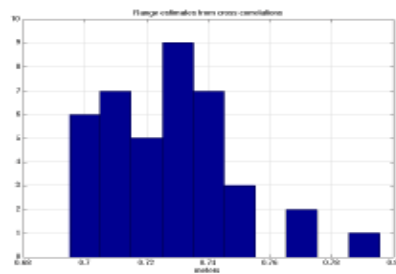




**Velocity estimates agree with exact values  
within experimental error**



**Range estimates from cross correlations are  
reasonable**





### Conclusions

---

- Erskine's velocimetry technique has been demonstrated for coherent broad band and narrow band acoustic pulses
- Accuracy is increased with longer initial delay times or finer time discretization ( $\Delta t / t_1 \ll 1$ )

## Grating Lobe Reduction in Large Element, 2-Dimensional Phased Arrays

*Jan-Ulco A. Kluiwstra*



# Denoising Data using Wavelet Based Methods

UCRL-VG-136075

*by Chuck Baldwin and Chandrika Kamath*

An important and highly researched problem in many fields of physics, engineering, and mathematics is the extraction of true data from data that has been contaminated by additive Gaussian noise. The use of wavelet coefficient thresholding has been an active topic of research for de-noising signal and image data and has proven to be a useful technique in certain applications. We will motivate this talk by presenting an application of data mining using astronomical data-sets and describe how the underlying noise gives rise to difficulties in feature extraction. We will then present some of the research into data de-noising through the use of orthogonal wavelet transforms: including applicable noise models, issues of wavelet basis selection and level of multi-resolution decomposition, thresholding policies and the various techniques for threshold selection. Finally, we will conclude by presenting some results of the wavelet de-noising process on the previously mentioned data-set and what we have learned from the work.

Key Words: Wavelet Transform, Image Denoising, Thresholding

## Denoising Data Using Wavelet Based Methods

**Chuck Baldwin, Chandrika Kamath**

*Center for Applied Scientific Computing  
Lawrence Livermore National Laboratory*



## The FIRST survey

- We are working with scientists from the FIRST (Faint Images of the Radio Sky at Twenty-cm) project to automatically detect radio-emitting galaxies with particular morphologies. The data are from radio interferometric observations at the NRAO Very Large Array (VLA) in Socorro NM. Currently, around 400,000 sources contained in approximately 100 GB of data have been collected.
- The data exhibits Gaussian noise as well as artifacts from the reconstruction of the signal. It is important to extract as much of the true data as possible since the processed data will be used in later pattern recognition techniques. The success of those techniques will depend on the quality of the denoised data.

## The FIRST data

Some of the details of the observations are :

- Over 10,000 square degrees of the North Galactic Cap.
- 3-minute snapshots are acquired covering a hexagonal grid.
- The data are edited, self-calibrated, mapped, and CLEANed using AIPS.
- A final atlas of maps is produced by coadding the twelve images adjacent to each pointing center.

Some of the details of the data are :

- Maps have 1: 8" pixels with a resolution of 5" and a typical rms of 0: 15 mJy.
- The noise in the coadded maps varies by only 15% from the best to the worst places in the maps, except in the vicinity of bright sources (> 100 mJy) where sidelobes can lead to an increased noise level.
- At the 1 mJy source detection threshold, there are ~ 90 sources per square degree, ~ 35% of which have resolved structure on scales from 230".



## Introduction

Suppose we have  $N$  true data values  $\{f(\chi_i), i = 1, \dots, N\}$  which are collected with noise introduced

$$\begin{aligned} Y_i &= f(\chi_i) + \epsilon_i \\ &= f_i + \epsilon_i \end{aligned}$$

where

$$\epsilon \sim N(0, \sigma^2)$$

is the noise component.

The goal is to reduce the effects of the noise so that the true data values can be recovered. As can be expected, the “quality” of the recovered true data will be directly related to how well the noise can be estimated and modeled, i.e. reasonable SNR (Signal to Noise Ratio).

## The Wavelet Decomposition I

Let  $W$  represent the orthogonal matrix associated with the orthonormal wavelet system of choice. Suppose

$$Y = (Y_1, Y_2, \dots, Y_N)^T$$

is a vector associated with the noisy data values and set

$$w = WY = W(f + \epsilon).$$

The “empirical” wavelet coefficients

$$w = (w_{0,0}, w_{1,0}, w_{1,1}, \dots, \theta_{J-1,2^{J-1}})^T$$

are a noisy estimate for the true wavelet coefficients

$$\theta = Wf = (\theta_{0,0}, \theta_{1,0}, \theta_{1,1}, \dots, \theta_{J-1,2^{J-1}})^T$$

converging at the parametric rate

$$w_{j,k} - \theta_{j,k} \sim N(0, \sigma^2).$$



## The Wavelet Decomposition II

The relationship between the coefficients is expressed as

$$w_{j,k} = \theta_{j,k} + z_{j,k},$$

where the set of  $z_{j,k}$  are a set of (unobservable) independent  $N(0, \sigma^2)$  random variables.

Each empirical wavelet coefficient consists of a certain amount of noise but relatively few contain significant signal. This leads to a heuristic of how the true data could be extracted from the noise.

1. Separate coefficients which contain significant signal from those that contain mostly noise.
2. Once the significant coefficients are found, attempt to remove the noise from the empirical coefficients by dropping (or shrinking) the coefficients that contain mostly noise.

## Wavelet Thresholding I

These two points are justified by noting :

- Large “true” coefficients  $\theta_{j,k}$  will typically have large corresponding empirical coefficients  $w_{j,k}$ .
- The “noisy” coefficients  $z_{j,k}$  contribute little to the “true” coefficients and can typically be ignored.

This heuristic yields the idea of thresholding the empirical coefficients.



## Wavelet Thresholding II

Thresholding the empirical coefficients yields an estimate of the true coefficients

$$\tilde{\boldsymbol{\theta}} = (\tilde{\theta}_{0,0}, \tilde{\theta}_{1,0}, \tilde{\theta}_{1,1}, \dots, \tilde{\theta}_{J-1,2^{J-1}-1})^T$$

and an approximation to the true data

$$\tilde{f} = W^T \tilde{\boldsymbol{\theta}}$$

The thresholding estimator of the true coefficients is written as

$$\tilde{\theta}_{j,k} = \sigma \delta_{\lambda_k} \frac{(w_{j,k})}{\sigma}.$$

The non-linear threshold function  $\delta_{\lambda_k}()$  has the task of making the determination of the significance of a empirical coefficient as well as modifying the coefficient itself.

## Estimating the Variance

The variance  $\sigma^2$  required in the threshold estimator is usually not known so an estimate is required. The median of absolute deviation (MAD) of a sample is given as

$$\text{MAD}(Z_1, Z_2, \dots, Z_N) = \text{median}_k(|Z_k - \text{median}_k(Z_k)|).$$

This quantity applied to the highest level coefficients is usually used as an estimate  $\sigma^2$ .

$$\hat{\sigma}^2 = \frac{\text{MAD}(w_{J-1,k})}{0.6745}$$

The mean of absolute deviation could also be used as an estimator for  $\sigma^2$ .



## Denoising Process

The denoising process is given by the following procedure

1. Compute the wavelet transform  $w = \mathcal{W}Y$ .
2. Estimate the variance  $\sigma^2$  by  $\hat{\sigma}^2$ .
3. Apply a non-linear threshold function  $\delta_{\lambda_k}()$  to the empirical coefficients  $\bar{\theta}_{j,k} = \sigma \delta_{\lambda_k}(\frac{w_{j,k}}{\sigma})$ .
4. Compute the inverse wavelet transform  $\bar{f} = \mathcal{W}^T \bar{\theta}$ .

The success will depend on how well the threshold function can find and remove the noise from the empirical coefficients.

## Thresholding Functions

Several choices for the threshold function  $\delta_{\lambda_k}$  are given in the literature.

The *hard thresholding function* is given by

$$\delta_{\lambda_k}^H(x) = \begin{cases} x & \text{if } |x| > \lambda_k \\ 0 & \text{if } |x| \leq \lambda_k \end{cases}$$

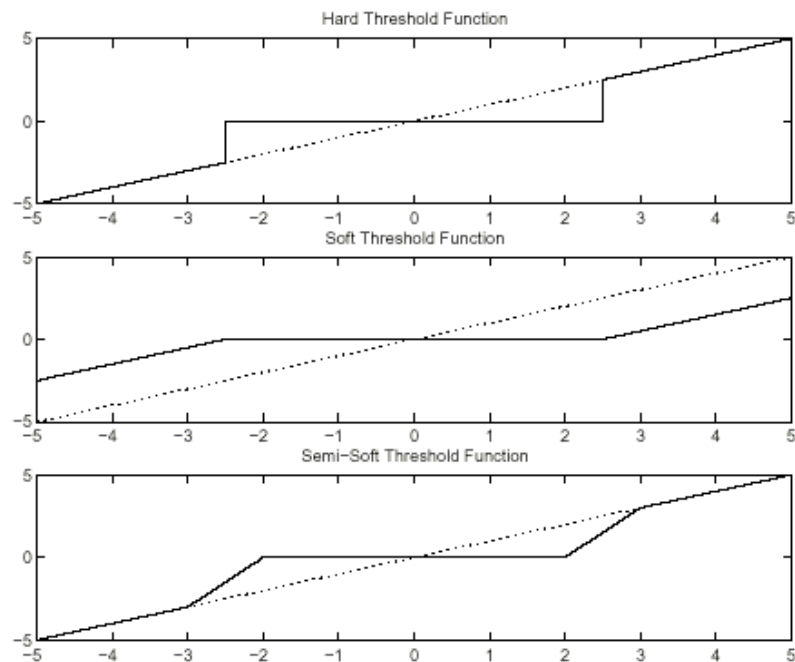
The *soft thresholding function* is given by

$$\delta_{\lambda_k}^S(x) = \begin{cases} x - \lambda_k & \text{if } x > \lambda_k \\ 0 & \text{if } |x| \leq \lambda_k \\ x + \lambda_k & \text{if } x < -\lambda_k \end{cases}$$

The *semi-soft thresholding function* is given by

$$\delta_{\lambda_{k,1}, \lambda_{k,2}}^{SS}(x) = \begin{cases} x & \text{if } x > \lambda_{k,2} \\ x - \lambda_k & \text{if } \lambda_{k,2} \geq x > \lambda_{k,1} \\ 0 & \text{if } |x| \leq \lambda_{k,1} \\ x + \lambda_k & \text{if } -\lambda_{k,2} \leq x < -\lambda_{k,1} \\ x & \text{if } x < -\lambda_{k,2} \end{cases}$$





## Methods of Threshold Selection

Many methodologies are in the literature on how to select thresholds :

- Manual thresholds
- Universal thresholds
- Mini-max thresholds, “RiskShrink”
- Steins Unbiased Risk Estimate, “SUREShrink”
- Cross validation
- Hypothesis testing ( $\chi^2$  or Brownian Sheet)
- Bayes rules



## Global Thresholding I

Global thresholding uses a single value of the threshold parameter and usually is applied to only one (or a few) level(s) of empirical wavelet coefficients.

$$\lambda_k = \begin{cases} \lambda & \text{if } k \geq \bar{J} \\ 0 & \text{if } k < \bar{J} \end{cases}$$

where  $\bar{J}$  is the lowest level to which thresholding will be applied.

Two ideas for choosing the threshold parameter  $\lambda$  are noted in the literature :

- *universal thresholding*
- *minimax thresholding*

## Global Thresholding II

The *universal thresholding* parameter is

$$\lambda = \sqrt{2 \log N} \sigma$$

The *universal threshold* depends only on the sample size  $N$ . It is derived from probabilistic arguments.

The *minimax thresholding* parameter is a constant of the form

$$\lambda = c(N) \sigma$$

The *minimax threshold* depends on the sample size  $N$  and the type of threshold function used. It is derived from minimizing the constant term in an upper bound of  $L^2$  risk.



## Level Dependent Thresholding

The idea of *universal thresholding* can be extended to level dependent thresholding by the following

$$\lambda_k = \sqrt{2 \log N} \sigma_k$$

where  $\sigma_k$  is the variance of the noise in the level  $k$  coefficients. Using the previous MAD variance estimator this can be stated as

$$\lambda_k = \sqrt{2 \log N} \frac{\text{MAD}(w_{J-1,k})}{0.6745}$$

## SURE Thresholding I

The least  $L^2$  risk of an estimator  $\tilde{f}$  to  $f$  is given by

$$R(\tilde{f}, f) = \mathbf{E} \left[ \frac{1}{n} \sum_{i=1}^n (\tilde{f}_i - f_i)^2 \right].$$

By Parseval's identity this yields

$$R(\tilde{f}, f) \propto \mathbf{E} \left[ \frac{1}{n} \sum_i \sum_j (\tilde{\theta}_{j,k} - \theta_{j,k})^2 \right].$$

So that minimizing  $\ell^2$  risk in the wavelet domain will automatically minimize  $L^2$  risk in the original domain.

This leads to the use of Steins Unbiased Risk Estimator (SURE) for threshold selection.



## SURE Thresholding II

The  $l^2$  risk can be estimated unbiasedly (for the soft threshold function) for a given  $t$  and  $x = (x_1, \dots, x_d)^T$  by

$$\text{SURE}(t, x) = -d + 2(\text{card}(\{k : |x_k| > t\})) + \sum_{k=1}^d \min(|x_k|, t),$$

where  $\text{card}()$  denotes the cardinality of a set. The threshold level  $\lambda$  is set so as to minimize the estimate of risk for given data  $(x_1, \dots, x_d)^T$ .

$$\lambda = \arg \min_{t \geq 0} \text{SURE}(t, x)$$

The SURE criteria is usually used to select a threshold for each level of coefficients

$$\lambda_k = \arg \min_{t \geq 0} \text{SURE}(t, \{w_{j,k}\}_j)$$

Implementation typically involves sorting the coefficients and performing a calculation on each value  $|w_{j,k}|$ .

## Measuring Quality I

In order to quantify how much noise has been removed from the true data and how much the true data has been altered a few measures are needed.

Given a collection of data  $y = (y_1, \dots, y_N)^T$  the *energy* or  $l^2$  norm is defined as

$$\|y\|_2^2 = \sum_{i=1}^N |y_i|^2.$$

and gives an idea as to the "content" of the data.



## Measuring Quality II

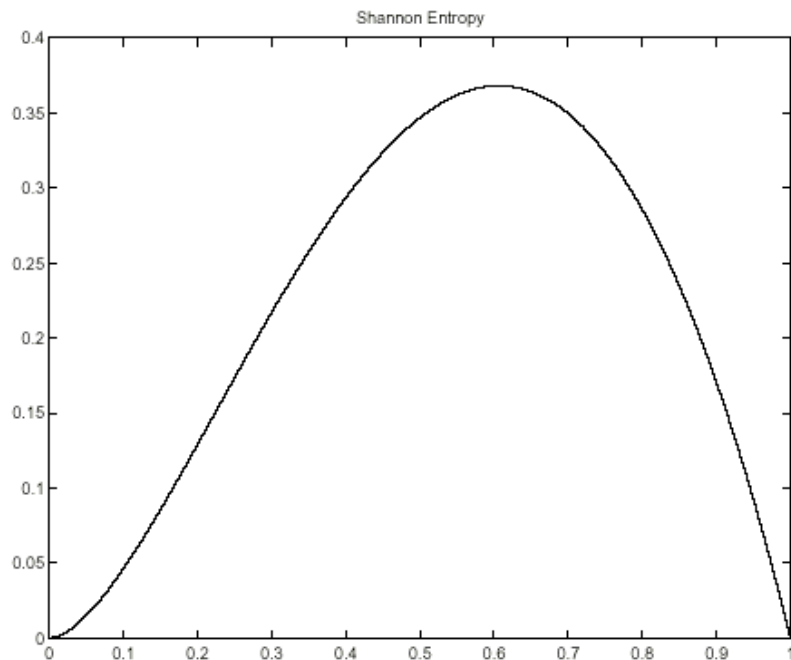
Given a discrete probability  $p = (p_1, \dots, p_M)^T$  the *log entropy* is defined as

$$E^L(p) = - \sum_{i=1}^M \begin{cases} \log(p_i^2) & \text{if } p_i \neq 0 \\ 0 & \text{if } p_i = 0 \end{cases}$$

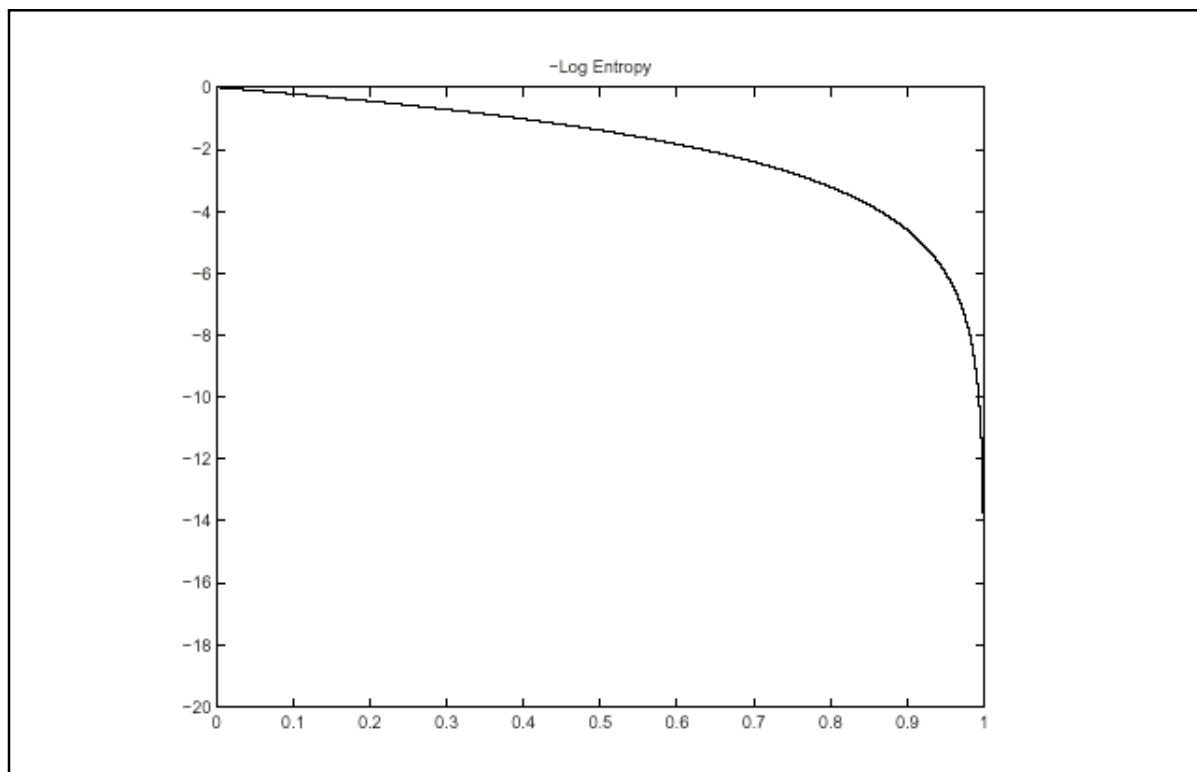
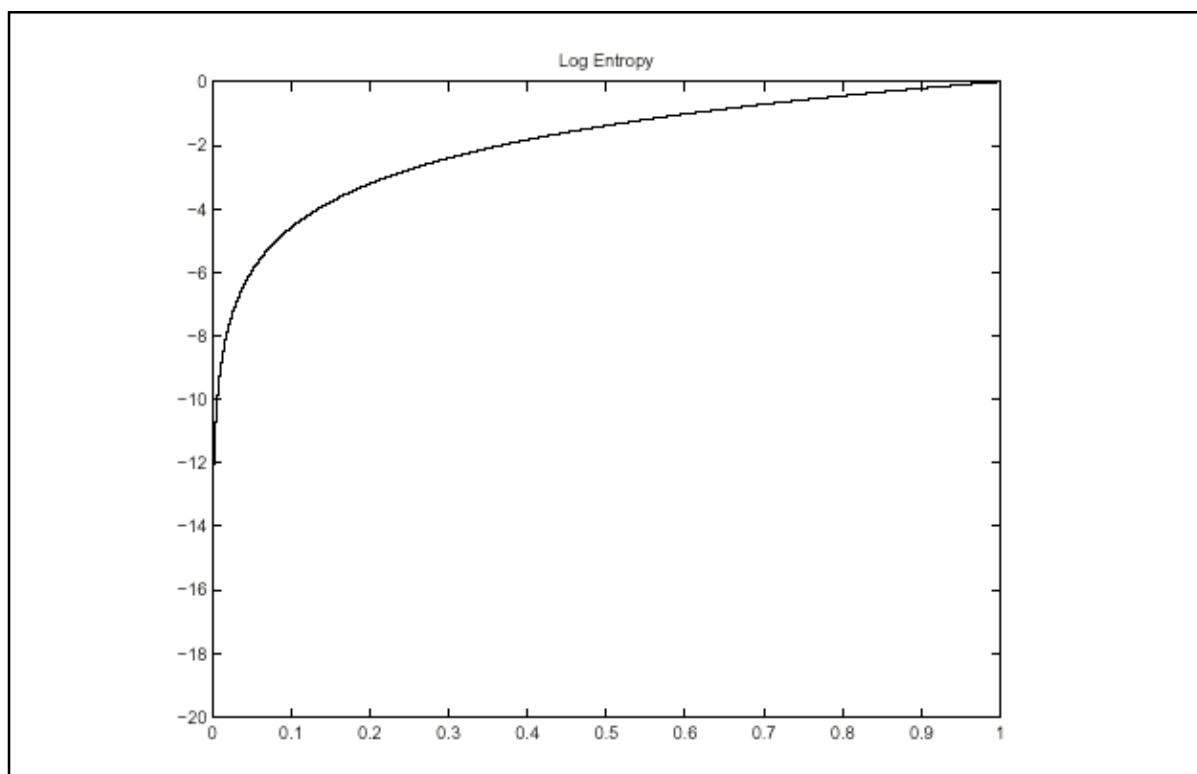
while the *Shannon entropy* is defined as

$$E^S(p) = - \sum_{i=1}^M \begin{cases} p_i^2 \log(p_i^2) & \text{if } p_i \neq 0 \\ 0 & \text{if } p_i = 0 \end{cases}$$

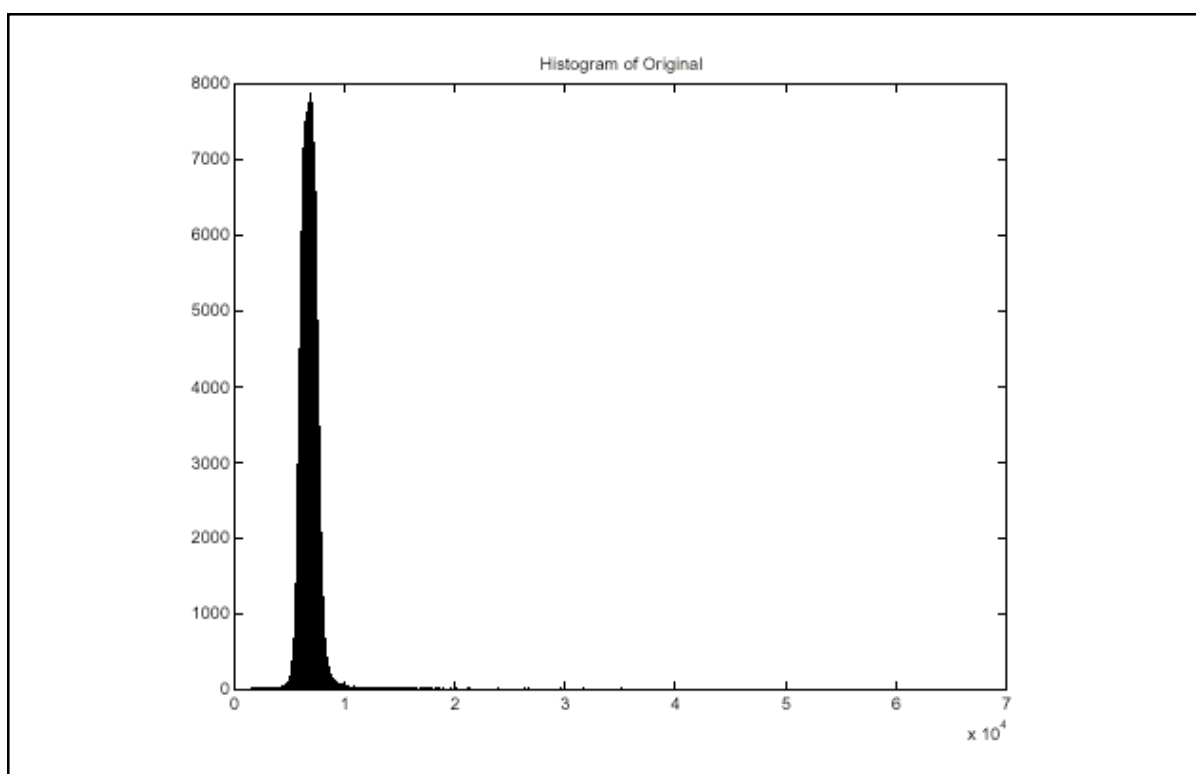
The *log entropy* weights the occurrence of low probability events less (or high probability events less with  $1-p$ ). The *Shannon entropy* weights the occurrence of low and high probability events less.







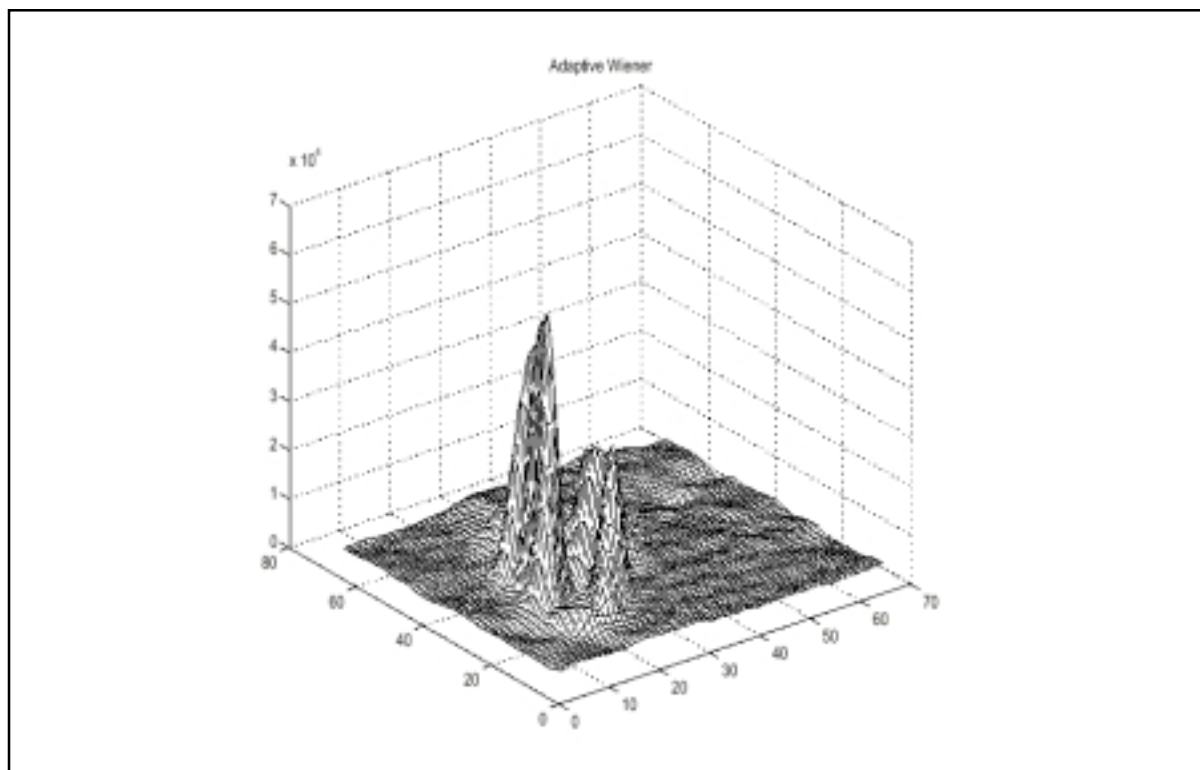
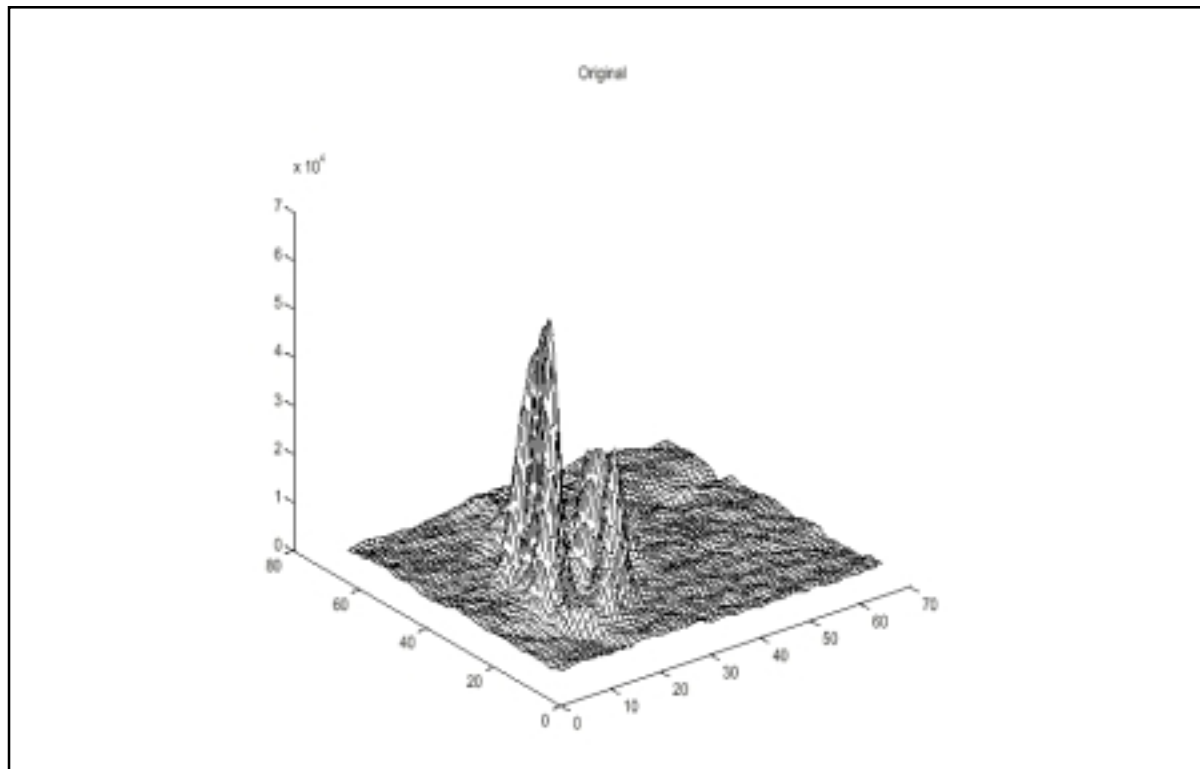




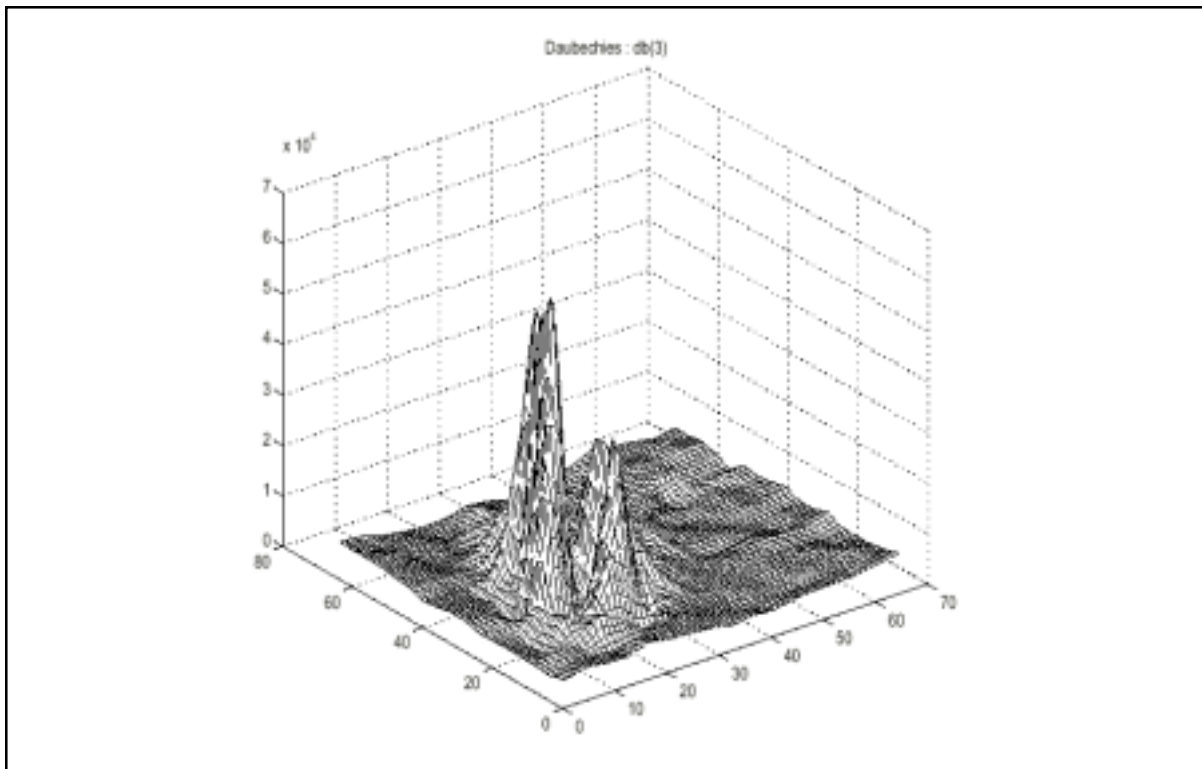
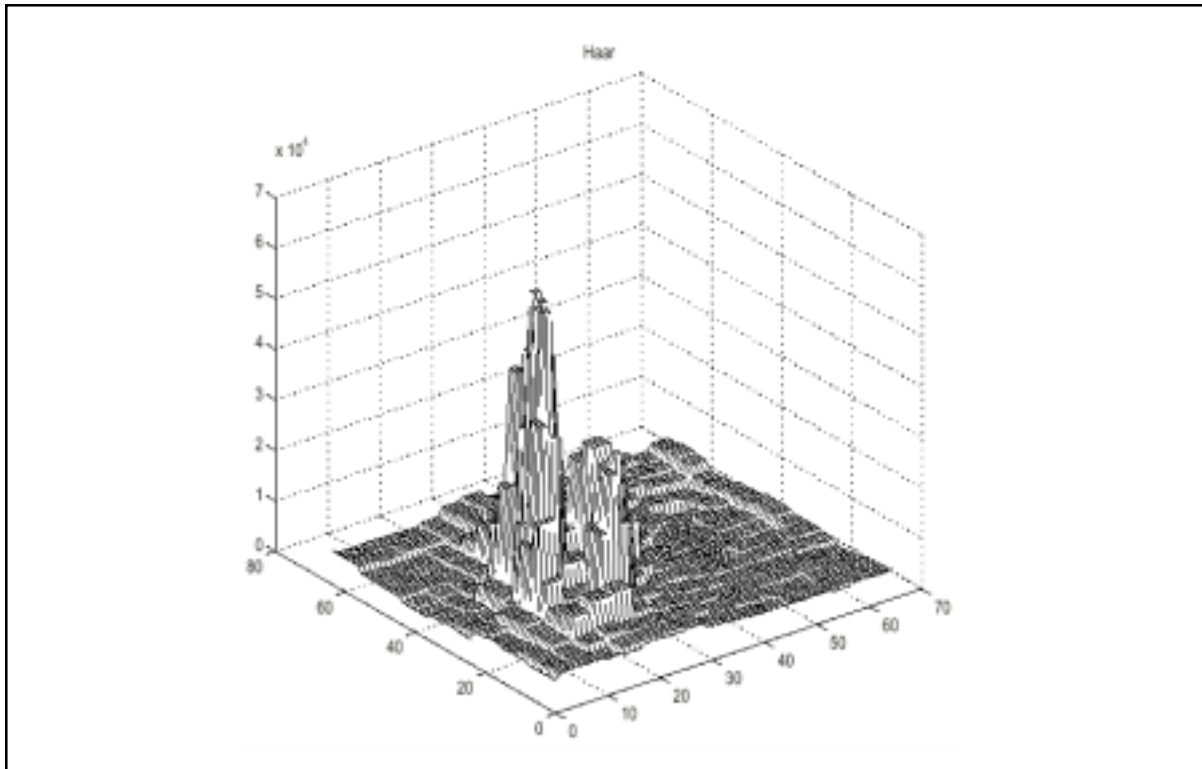
## Some Results

Data	$\hat{P}^2$ Energy	Entropy( $\hat{P}$ )	-Log Entropy( $\hat{P}$ )	-Log Entropy( $\hat{1} - \hat{P}$ )
orig	3.553270e+06	1.682872e-01	1.118174e+04	2.023206e+00
Wiener	3.296713e+06	2.799354e-02	9.819083e+04	2.002953e+00
Haar	3.496713e+06	8.511445e-03	5.211110e+04	2.000581e+00
db(2)	4.898353e+06	6.658038e-03	1.516255e+05	2.000441e+00
db(2)	2.819627e+06	5.724413e-03	1.501419e+05	2.000270e+00
db(4)	3.290733e+06	7.066679e-03	1.516996e+05	2.000409e+00
db(3)	3.517732e+06	6.198906e-03	1.623406e+05	2.000401e+00
db(6)	3.240903e+06	6.053905e-03	1.678265e+05	2.000294e+00
coif(1)	3.580574e+06	7.383453e-03	1.498432e+05	2.000106e+00
coif(2)	3.445794e+06	6.494175e-03	1.518072e+05	2.000424e+00
bio(1,3)	3.254405e+06	7.585328e-03	6.488030e+04	2.000512e+00
bio(1,5)	3.534829e+06	6.899556e-03	5.678677e+04	2.000455e+00
bio(2,2)	3.693312e+06	7.113143e-03	1.492778e+05	2.000474e+00
bio(2,4)	3.241648e+06	6.113072e-03	1.688038e+05	2.000235e+00
bio(2,6)	3.433351e+06	6.856072e-03	1.492494e+05	2.000451e+00
bio(3,1)	4.750812e+06	2.055296e-03	3.442290e+05	2.000417e+00
bio(3,3)	3.429335e+06	5.678379e-03	1.767756e+05	2.000264e+00
bio(3,5)	3.528303e+06	6.058788e-03	1.653022e+05	2.000291e+00
bio(4,4)	3.280159e+06	6.002278e-03	1.767176e+05	2.000387e+00

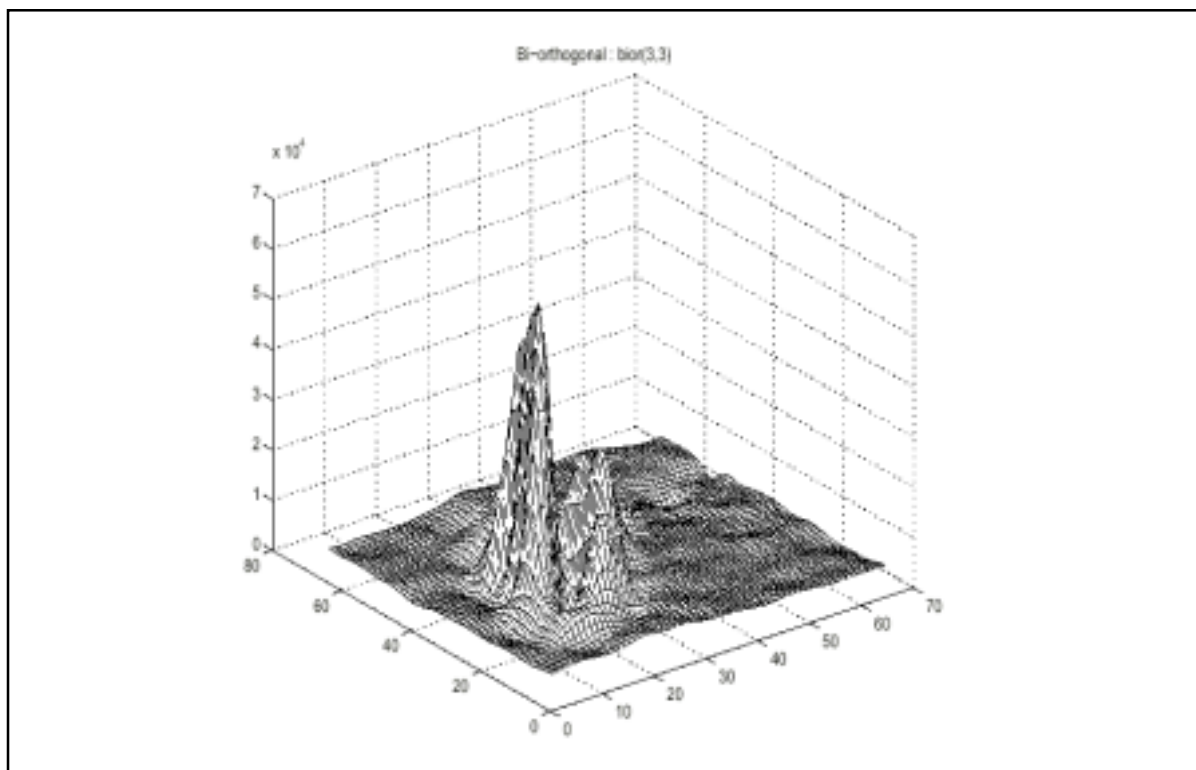
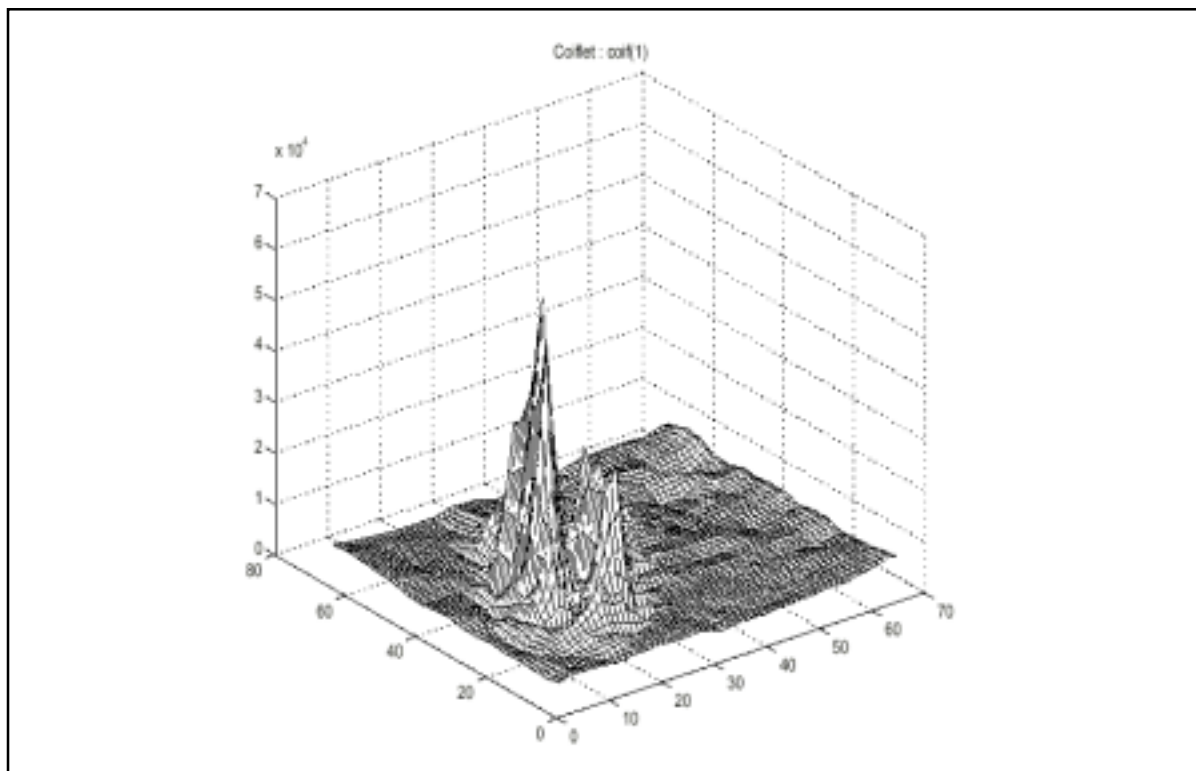














## More Results

Data	$l^2$ Energy	Entropy(p)	Mean	Std. Dev.
orig	3.559270e+06	1.682872e-01	6.847672e+03	1.198133e+03
Expert	2.612073e+06	6.747421e-03	4.952996e+03	1.222798e+03
SURE	2.880112e+06	6.296412e-03	5.485141e+03	1.247522e+03
Sqt2Log	3.554193e+06	1.841299e-02	6.837687e+03	1.197662e+03
Mini-Max	3.555799e+06	2.473489e-02	6.840845e+03	1.197819e+03

## Observations

Good Points :

- Wavelet Denoising is an effective methodology for removing (Gaussian) noise.
- Many varieties of thresholding functions and selection techniques exist.
- Most of the techniques for thresholding are computational efficient and parallel.

Bad Points :

- Not suitable for all applications, not a black box denoiser.
- Difficult to understand how threshold selection effects denoising.
- Many of the newer threshold selection techniques need to be coded.



## **The Sapphire Team: Supporting a multi-disciplinary endeavor**

- Chandrika Kamath (Project Lead)
- Chuck Baldwin
- Erick Cantu-Paz
- Imola Fodor
- Roy Kamimura
- Nu Ai Tang

<http://www.llnl.gov/casc/sapphire>





## Hematoma Detection Using MIR

### CASIS Workshop

John Chang, PhD

David Scott



## The Need for the Hematoma Detector

### Problem:

- Hemorrhage of major cerebral vesiculature due to head trauma
- Life threat due to pressure on the brain and brain stem
- Closed head injuries cannot be positively diagnosed without CT
- Delay presentation of clear symptoms with associated increase in mortality

### Solution:

- Exploit LLNL Micropower Impulse Radar (MIR) technology for early screening
- Provide in field diagnosis and triage
  - Neurosurgical procedure preparation
  - Govern patient evacuation method
- Improve overall patient management
- Life saving diagnostic tool





## Available Technology



### About Micropower Impulse Radar

- low power - runs on 9 V. battery
- small and very lightweight
- low cost - MIR electronics can be mass produced for
- microwave radiation penetrates body relatively well
- safe - order of magnitude lower power than a cellular phone



## Radar Specifications



- Frequency range: 1.5 to 4GHz (3GHz center)
- Peak power: 1 Watt
  - (at 10cm, no gain: 800 microwatt/cm<sup>2</sup>)
- Average power: 400 microwatt
  - (at 10cm, no gain: 0.3 microwatt/cm<sup>2</sup>)
- Pulse repetition rate:  $2 \times 10^6$  pulses/sec
- Pulse duration:  $2 \times 10^{-10}$  sec (200ps)
- Energy per pulse:  $2 \times 10^{-10}$  Joules





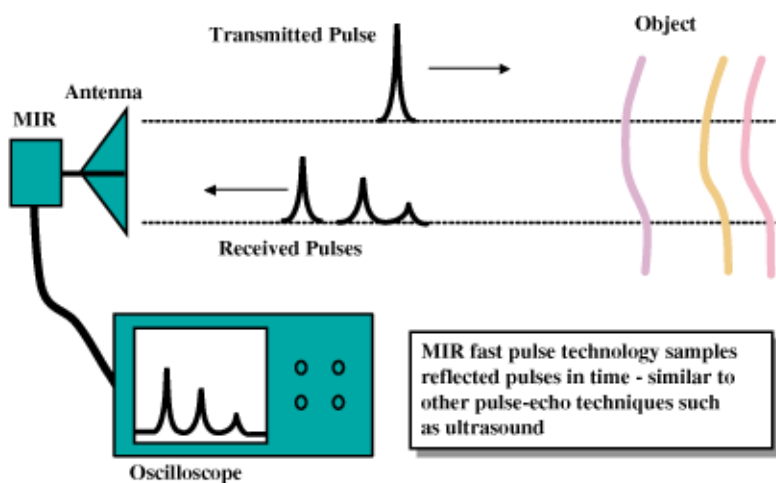
### Radar Specification Exposure Measurements (continued)



- Max antenna gain: 10
- Measurement distance from antenna: 10cm<sup>2</sup>
- Average power deposited: 0.003mW/cm<sup>2</sup>
- Peak power deposited: 8mW/cm<sup>2</sup>
- Peak energy per pulse: 0.0000016 *micro* J/cm<sup>2</sup>  
(1.6pJ/cm<sup>2</sup>)

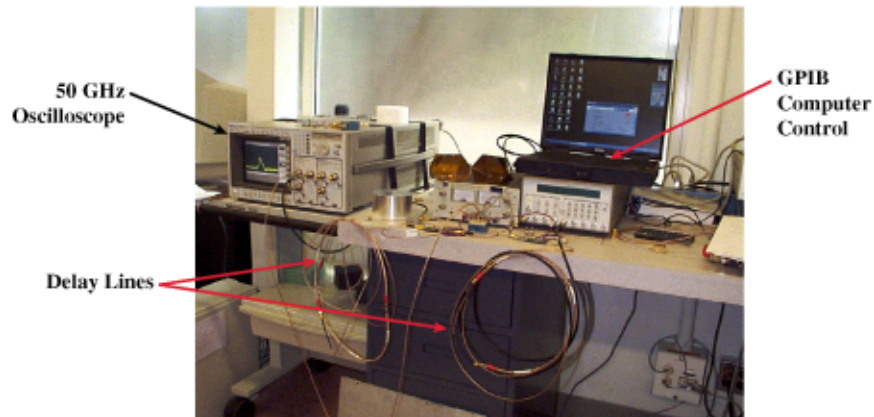


### How a “Rangefinder” Works





### Instrument based radar

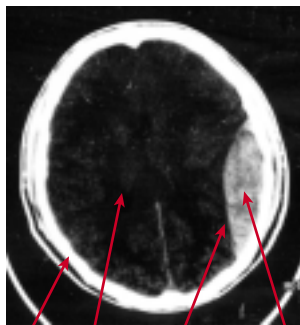


**MTP** MEDICAL TECHNOLOGY PROGRAM

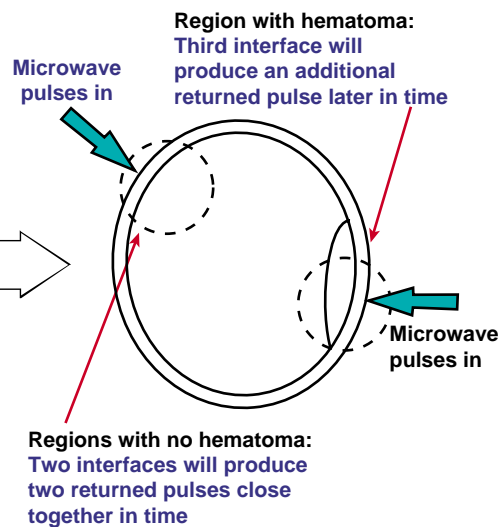
### The Hematoma Detector Records Returned Pulses from Dielectric Interfaces in the Head



CT slice image of a patient with a large epi-dural hematoma



Skull Brain Dura Fresh and clotting blood



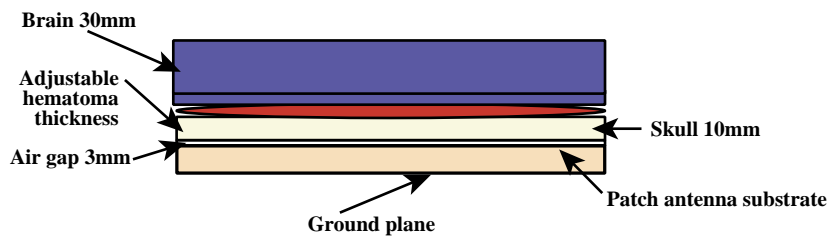
**MTP** MEDICAL TECHNOLOGY PROGRAM



## Results

- Numerical Model:

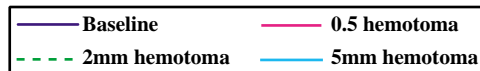
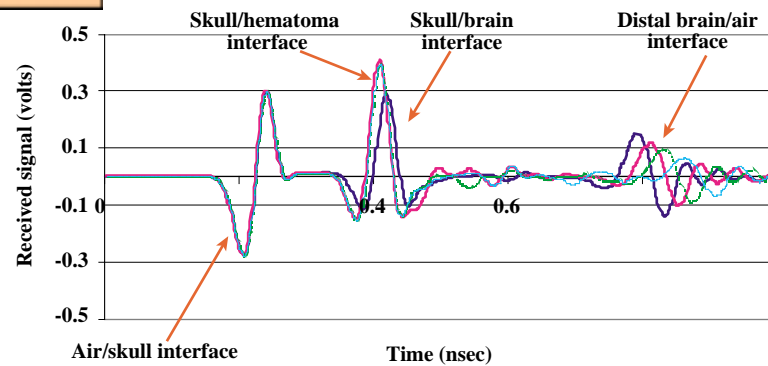
- Rectangular stratified slabs
- Patch antenna: coplanar rectangular, spiral, point source feed, dipole
- Transmitter/receiver distance is 30mm



**MTP** Medical Technology Programme

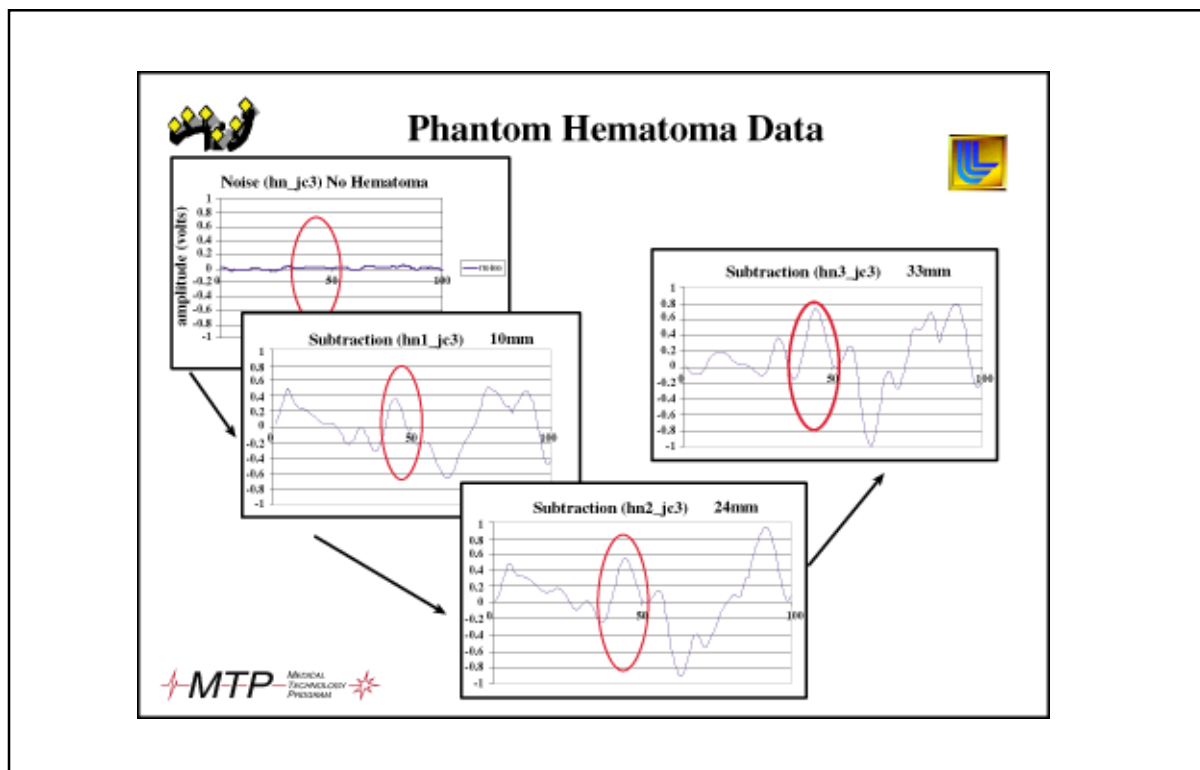
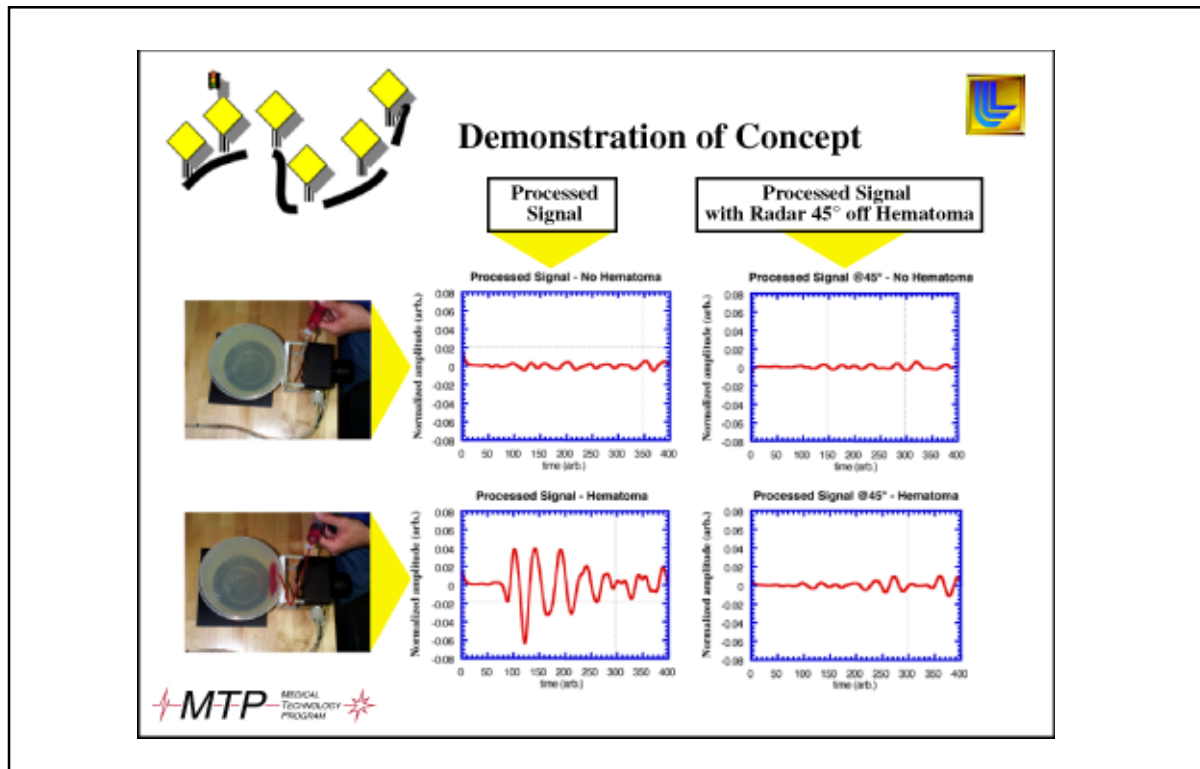
## Results

### Effect of hematoma presence



**MTP** Medical Technology Programme



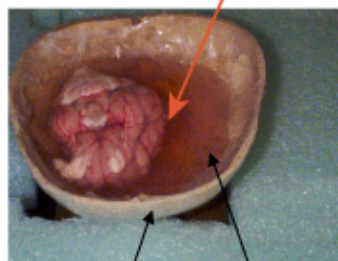




## Porcine brain experiment



### Hematoma phantom



Human skull

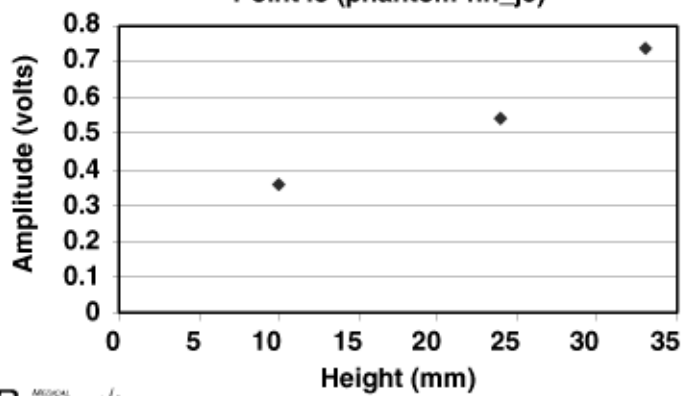
Saline



## Results



Point45 (phantom-hn\_jc)









---

# Imaging and Detection

---



# Towards Automated Optical Evaluation of Protein Crystals

UCRL 991112

*Pat Fitch, Brent Segelke, Bernhard Rupp  
Biology and Biotechnology Research Program  
Lawrence Livermore National Laboratory*

Preliminary results of our research towards developing an automated optical evaluation of protein crystals will be discussed. Several image processing approaches and algorithms are in the early stage of development and will be presented. Our algorithm development is performed on a desktop computer using MATLAB. The value of protein crystallization and the motivation for automating this part of the crystallization process is described below. If successful, a robust algorithm will allow more cost-effective high throughput protein crystallization to be implemented.

In the past few years there has been a tremendous increase in DNA sequence information available. In order to identify and understand the function of a gene, however, far more information than DNA sequence is required. Structural information about the protein that the DNA codes can provide significant information about function as well as insight into potential biochemical interactions (e.g., inhibitors), that can be used to modify function.

X-ray diffraction measurements of protein crystals can provide excellent structural information with resolution to about an Angstrom. The formation of protein crystals involves an educated search process for appropriate experimental conditions [1]. Each of the different samples (in some cases the same protein, but different experimental conditions) is monitored roughly 30 times before crystallization is complete. This monitoring is currently accomplished with a light microscope and visual evaluation. The images are given a quality score [2] to assist in determining if the crystallization process should continue, be aborted or modified, and/or fed-back into the crystal growth conditions. As automation, sample tracking and high throughput protein expression systems are integrated, the optical evaluation of the intermediate protein crystals becomes a bottleneck to high throughput. For instance, a robot can prepare 1,000 protein samples a day; each sample is imaged about 30 times before completion in roughly two months. If the robot keeps the pipeline full, this requires on average 1/2 an image each day for each sample. So at two months there would be 60,000 protein samples in process requiring 30,000 images to be viewed and evaluated each day. This is more than an image every second assuming an 8 hour workday.

[1] <http://www-structure.llnl.gov/> Web site for the structural biology group in the Biology and Biotechnology Research Program that includes protein crystal information and a nice tutorial.

[2] C.W. Carter Jr. and C.W. Carter, Protein crystallization using incomplete factorial experiments, *Biological Chemistry*, Vol. 254, No. 23, Dec. 10, 1979, pp. 12219-12223.



# In-Situ Sensing of Nucleation and Crystal Growth of $\text{Cd}_{1-x}\text{Zn}_x\text{Te}$

*B. William Choi*

The nucleation and growth of  $\text{Cd}_{1-x}\text{Zn}_x\text{Te}$  crystals in a multi-zone vertical Bridgman growth furnace have been observed and measured using in-situ eddy current sensor techniques. A two-coil eddy current sensor measured coil impedance changes for multifrequency which were then interpreted using an electromagnetic finite element analysis. The sensor was used to characterize the initial melting of a charge and the subsequent nucleation of solid during solidification. Fully remelted in-situ compounded charges were exposed to significant melt superheating and were found to undergo large melt undercoolings (of up to  $20^\circ\text{C}$ ), spontaneous crystal nucleation and rapid solidification (velocities approaching  $60\text{mm/h}$  which was more than ten times the furnace translation rate). Post growth metallography revealed that about  $20\text{mm}$  of polycrystalline solid was formed in this way before recalescence arrested the solidification interface. In partially remelted charges neither undercooling nor unstable growth were observed. These results indicate that eddy current sensors can be used to monitor critical aspects of the vertical Bridgman crystal growth of semiconducting materials and may simplify the implementation of seeded crystal growth concepts in this, and other, semiconductor crystal growth processes.



## **Real-Time Signal Processing of the Nucleation and Early Vertical Bridgman Growth of $\text{CdZnTe}$**

by

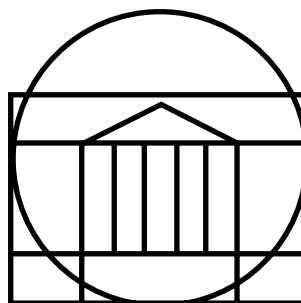
**ByoungWoo William Choi  
Hayden N. G. Wadley**





Presented at CASIS'S  
Signal & Imaging Sciences Workshop  
UC @ LLNL

Livermore, CA  
November 12, 1999



SCHOOL OF  
**ENGINEERING  
& APPLIED SCIENCE**

*Intelligent Processing of Materials*

*Dept. of Materials Science and Engineering*

**University of Virginia**

## **CdZnTe Ingot Grown by Vertical Bridgman Furnace JME~9035**

9035d.eps(tif)



**a) Boule Surface**

9035e.eps(pcx)

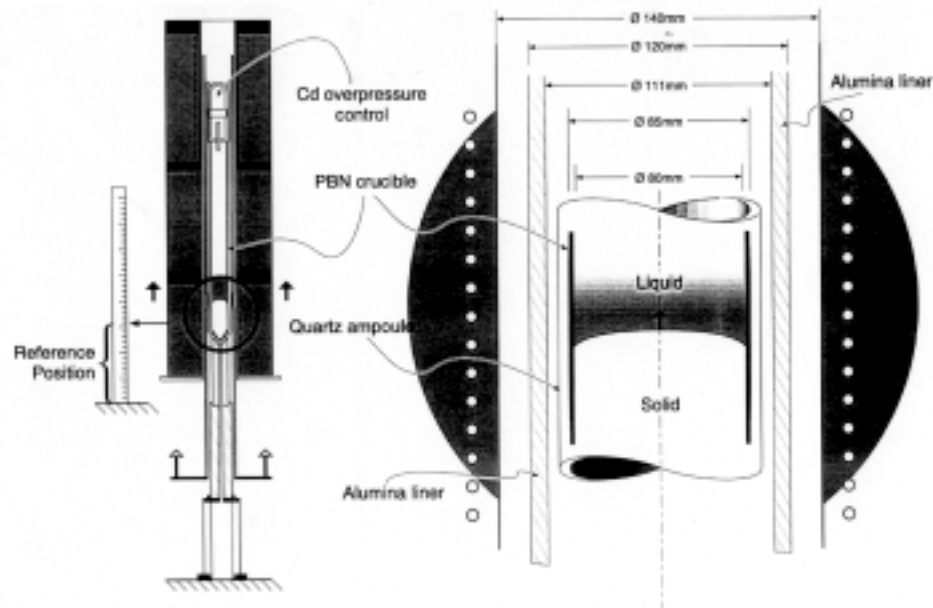


**b) Longitudinal Plane**

(photos by D. Bakken, JME, Spokane)



### Schematic Diagram of 17 Zone Vertical Bridgman Furnace



### CdZnTe Ingot Grown by Vertical Bridgman Furnace

- **Unseeded Directional Solidification**  
Cd: Te stoichiometric control by Cd overpressure techniques
- **Final Products**
  - Multigrain Ingot
  - Significant Zn Macrosegregation (i.e.  $0.02 \leq x \leq 0.07$ )
  - Frequent Tellurium Precipitation
  - High Dislocation Density etc.

The lowest stacking fault energy of all semiconducting compounds which crystallize in the sphalerite structure disappointingly low yield.



### CdZnTe Ingot Grown by Vertical Bridgman Furnace JME~9035

9035d.eps(tif)



a) Boule Surface

9035e.eps(pcx)



b) Longitudinal Plane

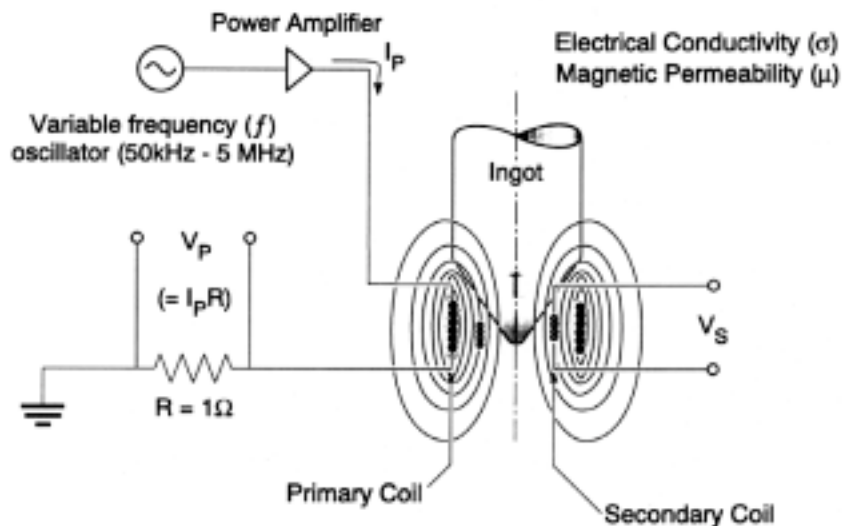
(photos by D. Bakken, JME, Spokane)

### CdZnTe Ingot Grown by Vertical Bridgman Furnace

- What should we know to improve this growth technology?
  - Melt Stoichiometry
  - Superheating and Undercooling
  - Nucleation/Growth
  - Solidification Velocity
  - Interface Shape
  - Temperature Gradients
  - Cooling Rate etc.



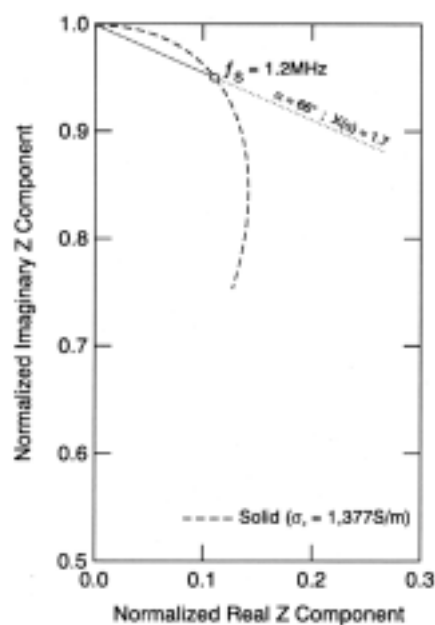
### Circuit Diagram of Two Coil Impedance Measurement System



**Measure:** Gain ( $G$ ) =  $V_s / V_p$   
Phase ( $\phi$ ) =  $\phi_s / \phi_p$

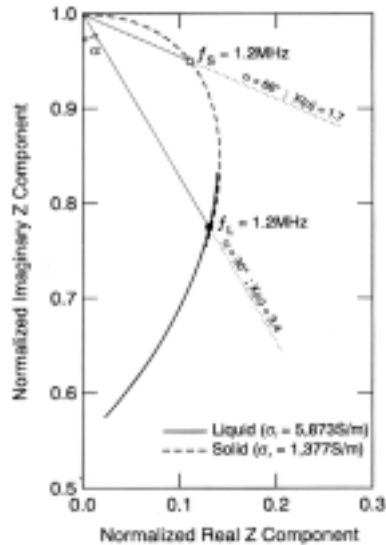
**Calculate:** Real  $Z = G/G_0 \sin(\phi - \phi_0)$   
Imaginary  $Z = G/G_0 \cos(\phi - \phi_0)$

### Calculation of Electrical Conductivity from Eddy Current Data





### Calculation of Electrical Conductivity from Eddy Current Data



$$\delta = \sqrt{\frac{2}{\omega \mu \sigma}}$$

$$X(\alpha) = \frac{R}{\sqrt{2\pi f \mu \sigma}}$$

where,  $R$  = radius of sample (m)  
 $f$  = frequency (Hz)  
 $\mu$  = magnetic permeability (H/m)  
 $\sigma$  = electrical conductivity (S/m)

Ref: *Introduction to Electromagnetic Nondestructive Test Methods*, H. L. Libby, John Wiley & Sons, Inc. (1971).

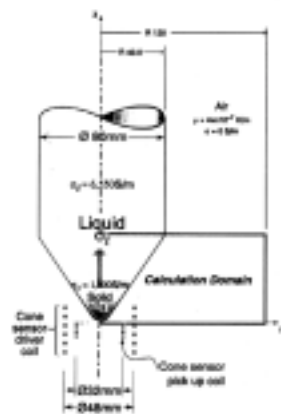
### Reference Silicon Cylinder Properties measured with a 4-point probe.

Ingot #	Conductivity (S/m) by 4 point probe	Conductivity (S/m) by Eddy Current	Frequency Range (kHz)
MN1-1861	1,471-1,149	$1,390 \pm 60$	800-1,200
MP0-5792	6,667-4,000	$5,810 \pm 170$	300-700
JME-31844	11,628-11,234	$11,700 \pm 170$	120-300

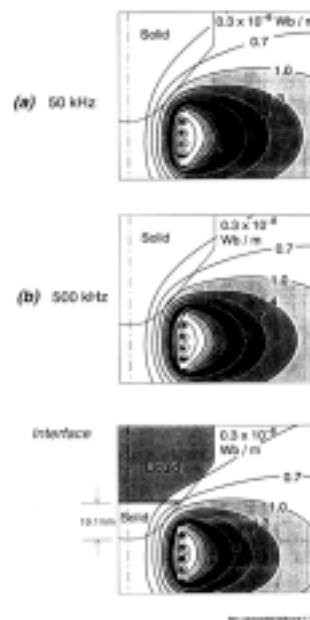
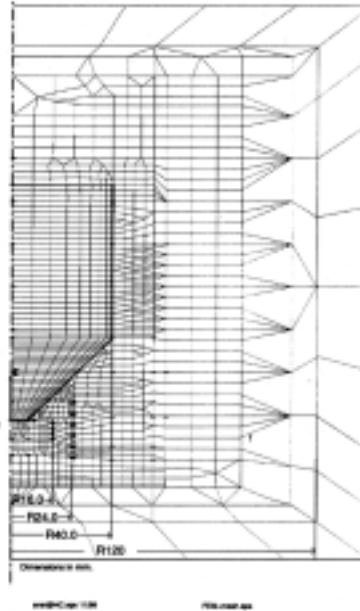


## FEM Geometry for Nucleation Sensor

### a) Finite Element Model Geometry

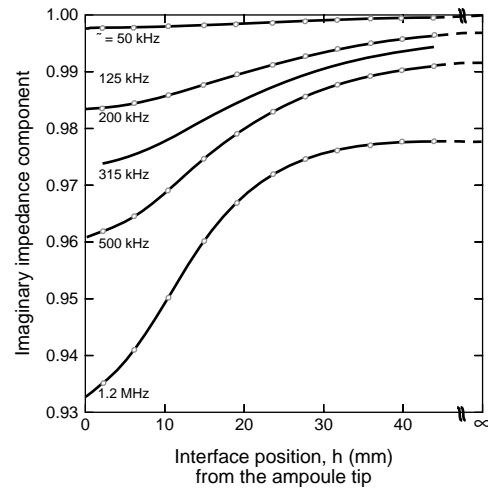


### b) Finite Element Mesh





## Nucleation Sensor



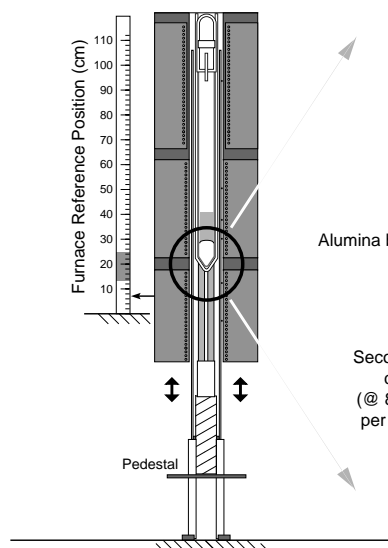
SCHOOL OF  
ENGINEERING  
& APPLIED SCIENCE

Intelligent Processing of Materials

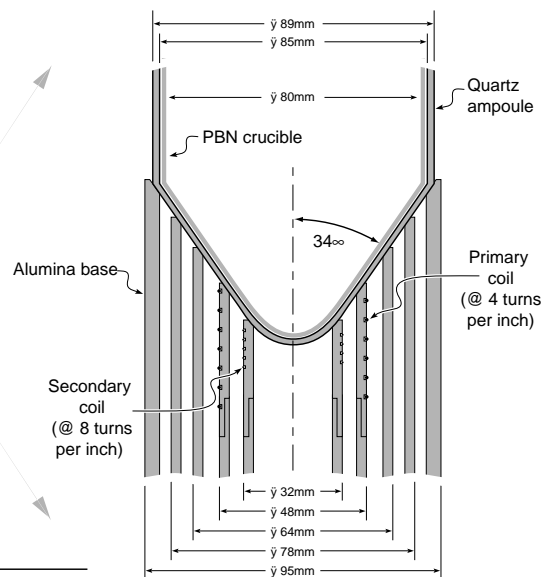
University of Virginia

## A Series of Eddy Current Sensing Configurations

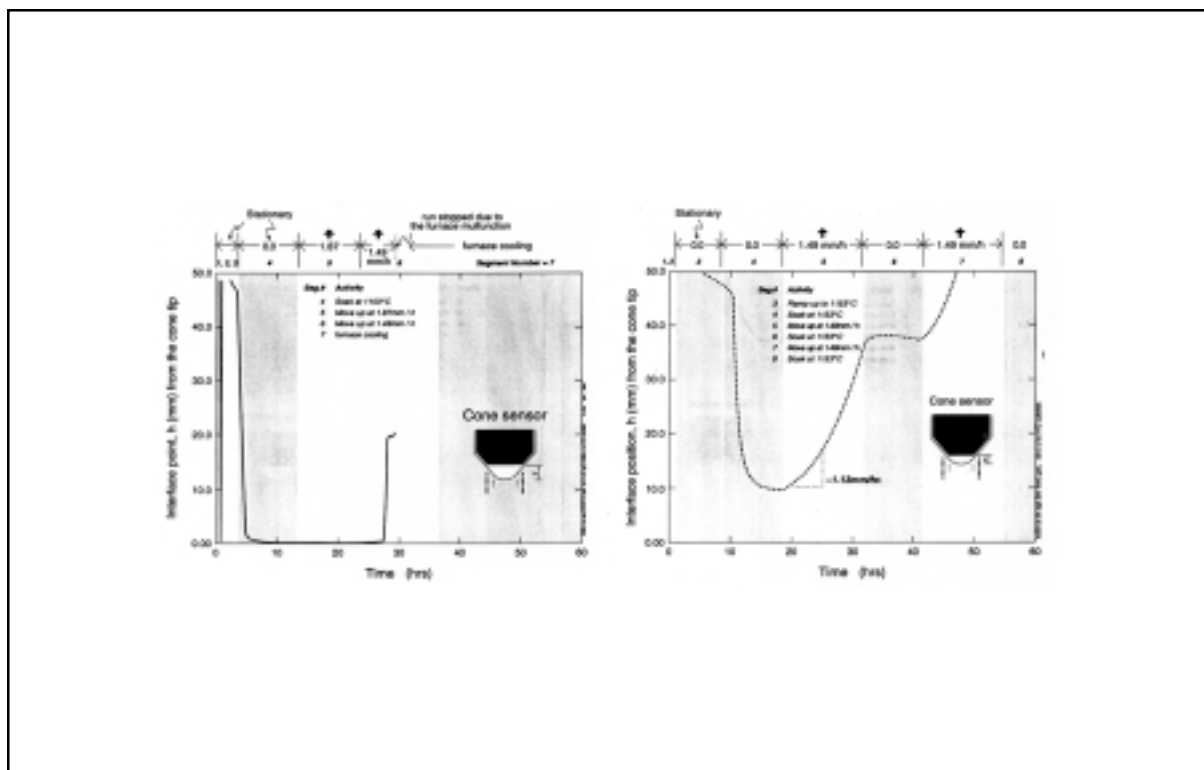
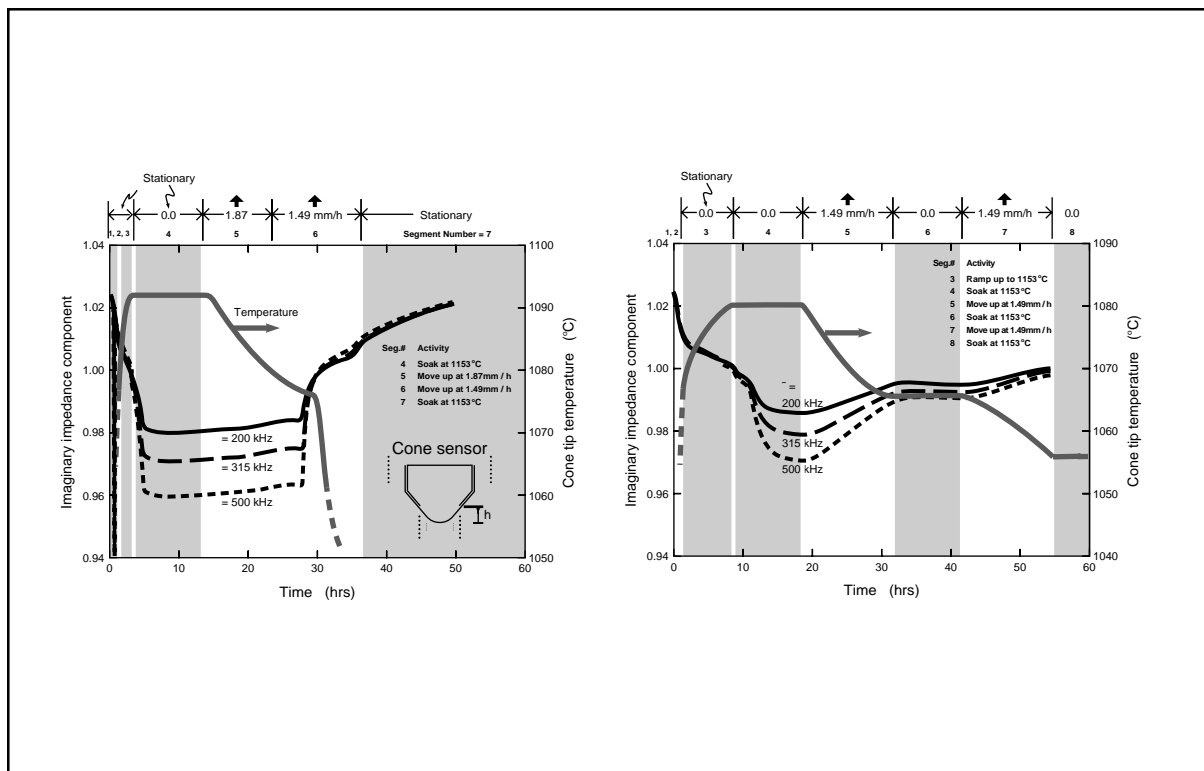
a) 17 zone Furnace



b) Cone Sensor

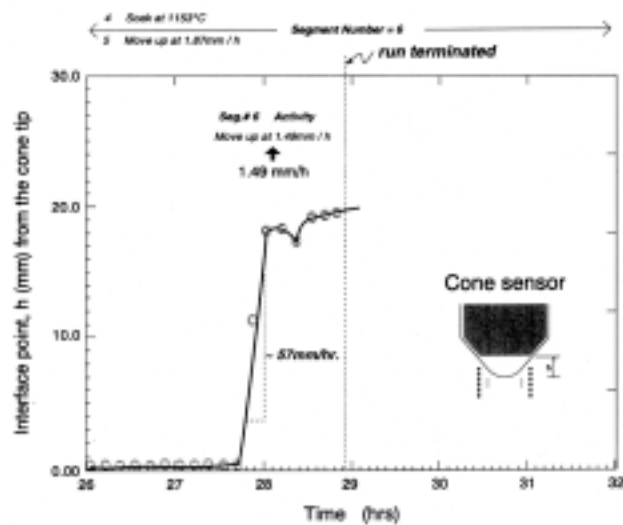




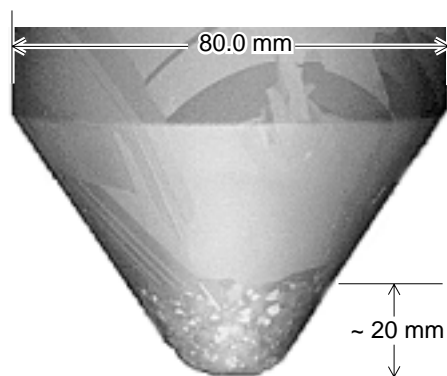




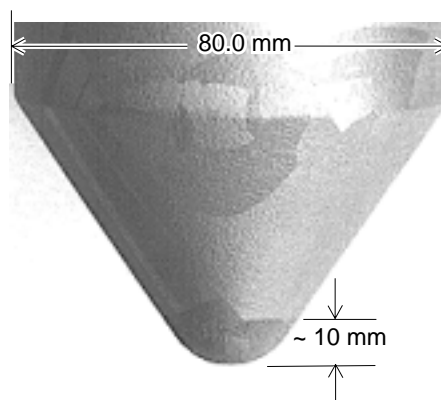
### Initial Nucleation/Growth (seg. #6)



### Photographs of the cone area showing the comparison of appearances between ingots grown



a) with completely melting state



b) with incomplete melting



### **In-situ Study of Crystal Growth**

---

- **Dual-coil encircling eddy current sensor system, non-contact, non-destructive**
  - **Melt Stoichiometry**
  - **Superheating and Undercooling**
  - **Nucleation/Growth**
  - **Solidification Velocity**
  - **Interface Shape**
  - **Temperature Gradients**
  - **Cooling Rate etc.**
- **Strategies**
  - **On-line sensing and feedback control**
  - **Seeded solidification processing**

Acknowledgements: UVa, JME, TI, and C. A.



## Phase Retrieval in X-ray Diffraction

**Abraham Szöke**

V Division

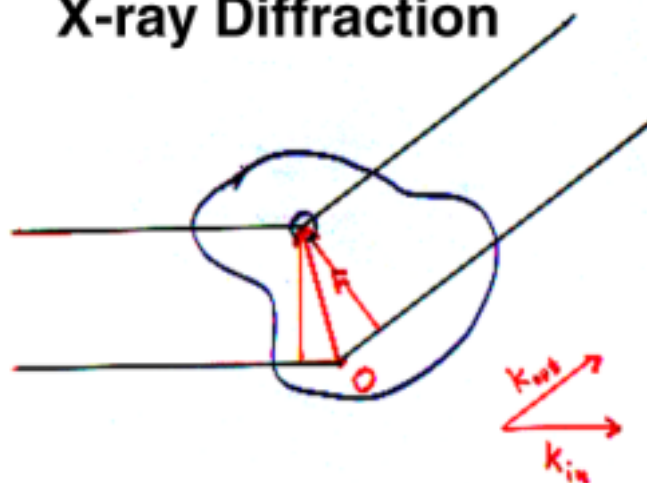
Lawrence Livermore National Laboratory  
Livermore, CA 94551

**November 11, 1999.**

“... lack of information cannot be  
remedied by mathematical trickery”

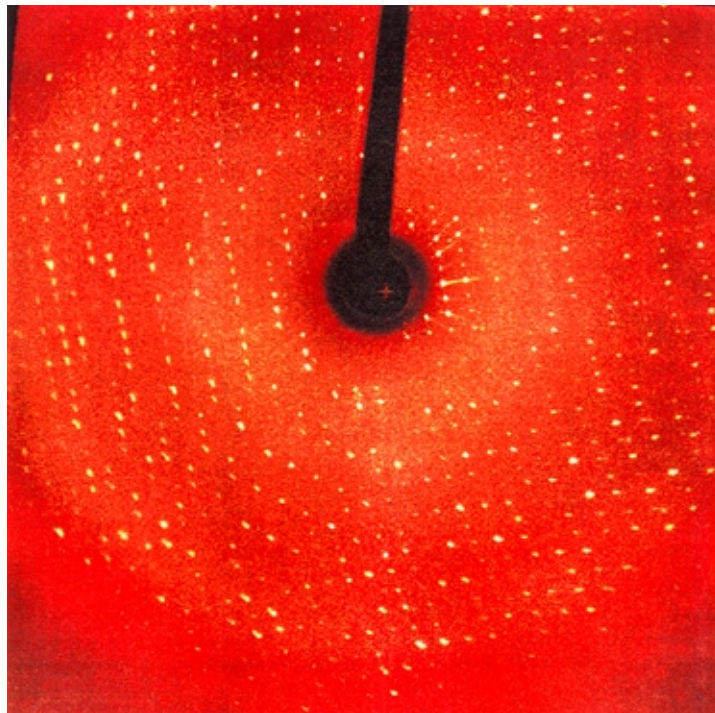
**Cornelius Lánzos**

### X-ray Diffraction



$$I(\mathbf{h}) = \int \rho(\mathbf{r}) \exp\{i(\mathbf{k}_{out} - \mathbf{k}_{in}) \cdot \mathbf{r}\} d\mathbf{r}$$





### The crystallographic "phase problem".

The incident X-ray beam is usually coherent over many unit cells. Then, the diffraction pattern of a crystal is that of a unit cell,

$$F(\mathbf{h}) = \int_{\text{unit cell}} \rho(\mathbf{r}) \exp(2\pi i \mathbf{h} \cdot \mathbf{r}) d\mathbf{r}.$$

It is sampled at integer values of the reciprocal lattice,  $\mathbf{h}$ .

Shannon's sampling theorem states that "a band limited function is completely determined by its sampled values, if the sampling is frequent enough."

Denote the Fourier transform of a function by  $f^\wedge(\mathbf{h})$ . Band limit means that  $|f^\wedge(\mathbf{h})| = 0$  if  $|\mathbf{h}| > \Omega$ . The theorem states that

$$f(\mathbf{r}) = \sum_n f(n\pi/\Omega) \frac{\sin(\Omega \mathbf{r} - n\pi)}{(\Omega \mathbf{r} - n\pi)}$$

More is true: undersampling gives freedom; critical sampling converges slowly; oversampling gives rapid convergence.



## The crystallographic "phase problem" (cont.)

*"... lack of information cannot be  
remedied by mathematical trickery."*

Cornelius Lanczos.

The Bragg condition results in integer values of  $h$ . It is exactly critical sampling.

From the point of view of information theory, half the information is missing.

Measured amplitudes of the reflections with arbitrary phases give an equally "valid" electron density.

More information is needed.

## More, independent information:

Experimental information:

MIR (Addition of the heavy atoms.)  
MAD (Anomalous scattering by some atoms.)  
NCS (Non crystallographic symmetry).  
More than one crystal form of the same molecule.

Prior knowledge:

Positivity of the electron density.  
Atomicity.  
Nearly uniform density in solvent regions.

Knowledge of part of the molecule:

(Called molecular replacement.)

Model building:

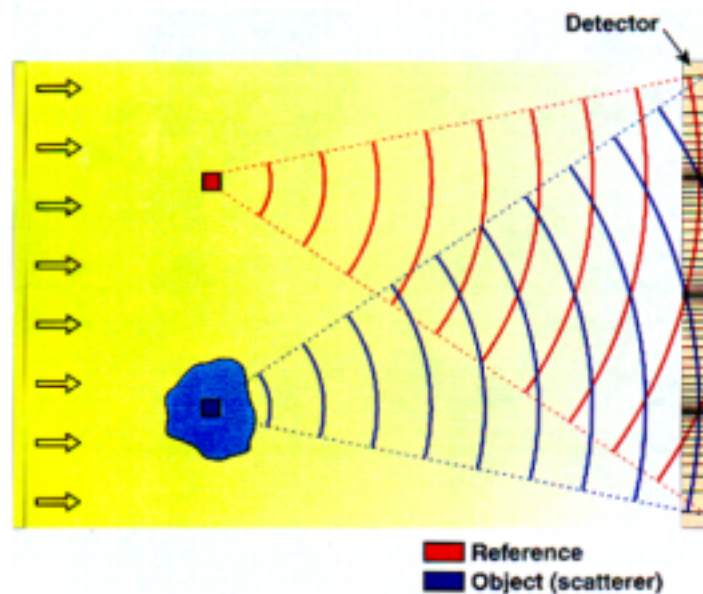
(Called refinement.)



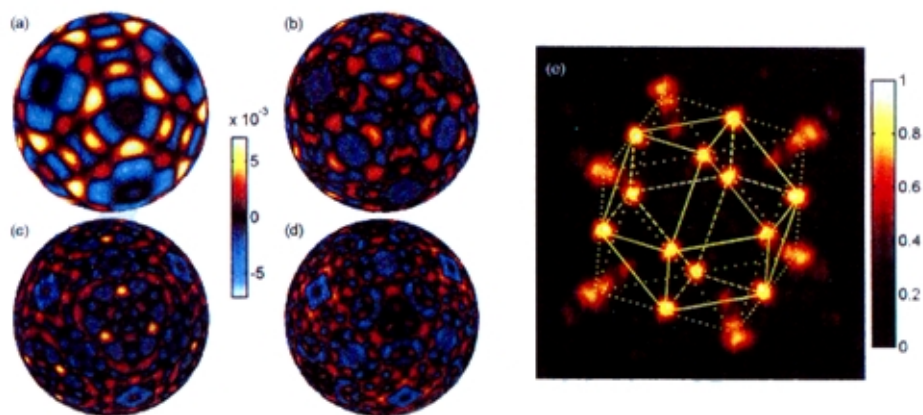
**Can we produce the entire  
diffraction pattern of a  
molecule from a crystal?**

Local reference holograms do that.

**Holographic Microscopy with a Local Reference**







### Diffraction of a crystal with two kinds of molecules.

We assume that the two kinds of molecules are similar and that they are thoroughly mixed.

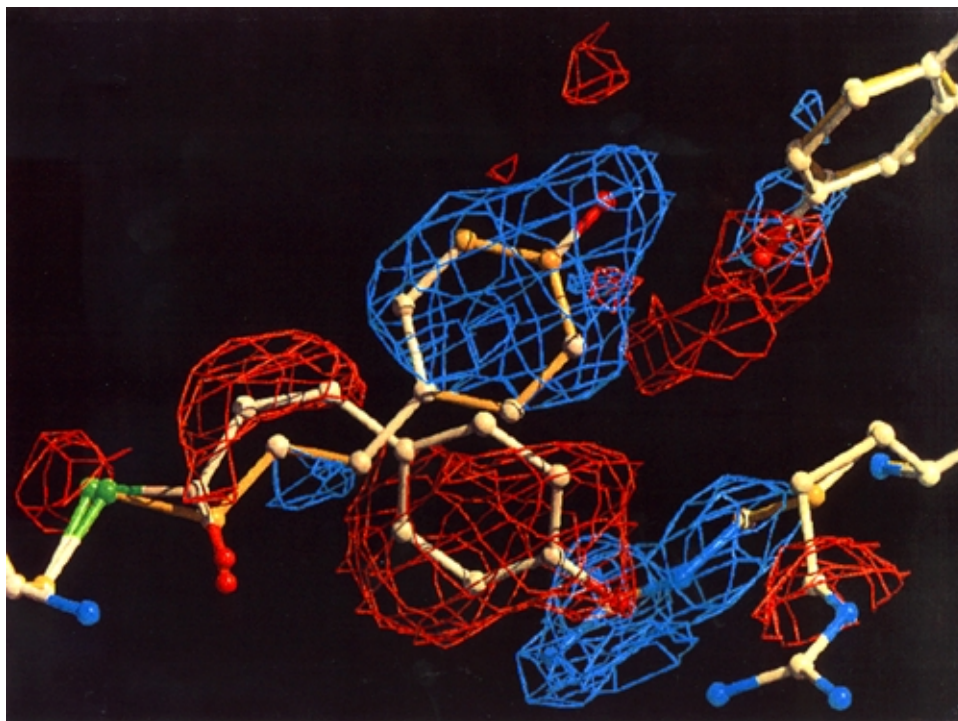
We can define an average density of the two molecules, and a difference density.

- The average density is periodic: it gives Bragg peaks.
- The difference density produces a continuous diffraction pattern of the difference molecule.

### Time resolved chemistry in crystals:

A reaction is initiated in a fraction of the molecules (e.g. by a short pulse laser.) A monochromatic X-ray pulse irradiates the crystal with a time delay. The diffraction pattern is recorded.





### Summary.

There is a close relation between X-ray crystallography and holography. (Gabor credited Bragg with sparking his ideas.)

The diffraction pattern of a crystal is that of a single unit cell, sampled at the reciprocal lattice points. There is not enough information in the diffraction pattern to reconstruct the electron density.

A continuous diffraction pattern of a molecule can be sampled finely enough to reconstruct it. (Except for the enantiomorph, the Babinet opposite and the position of the origin. This ambiguity is analogous to the dual image problem of holography.)

I have shown several ways to produce a continuous diffraction pattern of a molecule using a crystal:

- a local reference hologram,
- a disordered overlayer,
- a short coherence X-ray beam and
- a crystal lattice randomly occupied by two different molecules.







---

# Signal/Imaging Processing for NDE

---



# Nondestructive Characterization Center Overview

*Harry E. Martz, Jr.*  
*Lawrence Livermore National Laboratory*

The Nondestructive Characterization Center (NCC) supports initiatives that three to five years into the future will advance research, development, and cost effective application of nondestructive characterization science that enable successful LLNL and DOE programs. NCC plays a strategic and vital role researching and developing technologies to allow the Engineering Divisions to collaborate with and support LLNL and DOE programs. NCC pioneers nondestructive measurement science, develop and apply nondestructive measurement technology, and help transfer these advancements to LLNL and DOE programs, U.S. industry, universities, and other government agencies.

The NCC strategic mission objectives will advance core competencies and technologies in the following areas: (1) Quantitative NDE; (2) Fast Nanoscale Imaging; (3) Embedded in Design; (4) Multispectral; and (5) Portability. The NCC FY00 mission objectives advance core competencies and technologies vital to the following programs. Extraordinary laser systems, this includes advanced diagnostics, e.g., for fusion targets and automated in situ optics damage inspection. Weapon system performance, this includes 3D measurements over 15 cm to better than 25  $\mu\text{m}$ , advanced diagnostics, e.g., Advanced Radiography Campaign and ceramic properties. Science-based stockpile stewardship, specifically durability and life prediction. Nonproliferation, arms control and international security related to reducing the threat from weapons of mass destruction.



## Overview of the Nondestructive Characterization Center

prepared by

**Harry E. Martz, Jr.**

**Nondestructive Characterization Center  
Lawrence Livermore National Laboratory**



for  
**CASIS**

1999 Signal and Imaging Sciences Workshop  
Livermore, CA  
November 12, 1999

**The NCC's role is to R&D the core technologies  
and competencies to fulfill LLNL's mission**

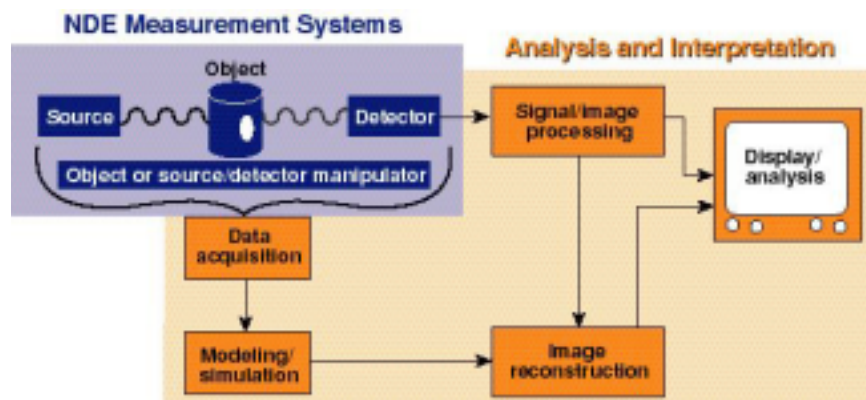


- **Vision**
  - To be the World's leader in the research, development, and cost-effective application of nondestructive measurement science that enable successful LLNL and DOE programs, and a national science and technology resource for U.S. industry, universities, and other government agencies
- **Mission**
  - Responsible for the vitality and growth of the NDC core technologies
  - Focuses and guides investments in enabling and supporting capabilities
  - Play a strategic and vital role researching and developing technologies to allow the Engineering Divisions to collaborate with and support LLNL and DOE programs
  - Pioneer nondestructive measurement science, research, develop and apply nondestructive measurement technology
  - Help Divisions transfer these advancements to LLNL and DOE programs, U.S. industry, universities, and other government agencies





## NDC system configurations are complex integrations of hardware and software



Multidisciplinary teams are essential in NDC R&D, system integration and data interpretation to solve challenging programmatic problems

hem 4/17/00



NCC-3

## The NCC strategic mission objectives will advance core competencies and technologies in five areas

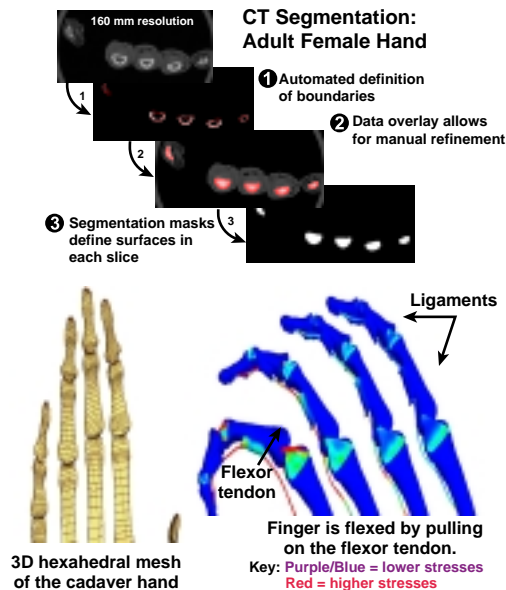


- **Quantitative NDE**—quantitative multi-dimensional nondestructive measurements to determine physical properties: geometry, density, elemental & molecular composition, thermal, structural, mechanical, stress state, radioactive assay, etc., in all three dimensions and as a function of time for the materials/objects we characterize.
- **Fast Nanoscale Imaging**—provide multiple programs with quantitative nanoscale multidimensional (>3D) imaging.
- **Embedded in Design**—unite currently very diverse and somewhat non-coupled technologies such as design and modeling, materials development, characterization and testing, process monitoring and control, reverse engineering, and reuse and waste management into a global package of integration and iteration.
- **Multispectral**—expand the spectral range of NDE imaging technologies and to increase the synergism between different methodologies.
- **Portability**—The objective of portability is to reduce current and new NDE systems to a size that is on the order of a small suitcase.





## Improved Prosthetic Implant Designs Using Computed Tomography to FEA Methods



- **Objectives**
  - New approach to evaluate and improve prosthetic joints before they are manufactured or surgically implanted
  - Conversion of CT data to FEW models
- **Challenges**
  - High definition 3D data of hard and soft human tissue
  - Automated segmentation w/o artifacts
  - CT and MRI data registration and fusion
- **Status**
  - Cadaver hands and knees have been CT scanned and reconstructed
  - The CT data have been segmented, surface rendered and meshed
  - FEA results have been used to predict failure and to improve implant designs

## Compelling NCC problems



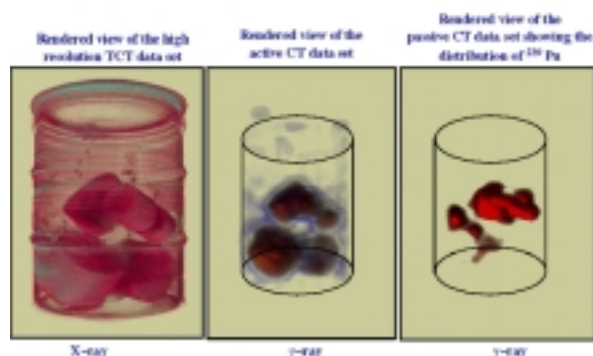
- **Measurement Systems**
  - See inside objects beyond what has been seen before
    - Improve spatial resolution down to nm scale
    - Increased penetration/contrast to see inside materials not accessible before
    - Temporal (fast) volumetric imaging
- **Analysis and Interpretation**
  - Multi-dimensional (>3D) signal and image processing
    - Speed, accuracy and noise reduction in processing
    - Data interpretation—We see more and more things; what does this mean to the customer
    - Fusion of different measurement systems data

Measurement systems, and analysis and interpretation must be developed jointly to get optimal synergy





## Automated multi-dimensional data acquisition, processing and interpretation is the world of the future



**Data Reduction:**  
3 grams  
weapons grade  
plutonium

**Decision:**  
Transuranic (not  
low-level) waste  
disposal required

Typically multi-dimensional data are required to produce a single number to make a crucial decision

**“GigaBytes to a Bit”**

hem 4/17/00



NCC-7

## Capabilities that would fundamentally alter the way our customer does business



- **Weapons surveillance**
  - Weapon system performance, this includes three-dimensional measurements over a large size with high spatial resolution, and advanced diagnostics, e.g., Advanced Radiography Campaign, and materials properties measurements
  - Science-based stockpile stewardship, specifically durability and life prediction
- **Precision diagnostics and devices**
  - Extraordinary laser systems, this includes advanced diagnostics, e.g., for fusion targets and automated in situ optics damage inspection
- **Arms control, nonproliferation, and international security related to reducing the threat from weapons of mass destruction**
  - Portable instruments // Imaging // Remote Sensing





## Details of NCC's future driving applications and collaboration with other Centers



- Team with PSEC & CμT in NIF target fabrication/characterization
  - Characterization/metrology
  - Diagnostics
- Team with CCE on NDE data to computational models & analysis
  - Simplify and automate the process to convert the NDE data to FEA meshes and data analysis
  - Adapt models to incorporate the new information provide by NDE
  - Research and develop life prediction
- Nano-scale characterization facility for materials characterization to predict life cycle in situ
  - Mechanical testing via micro- to nano-tensile testing, indentation, characterization in situ, etc.
  - Cycle back into design and fabrication

In order for this to be successful we also need to work very closely with Engineering's Divisions and LLNL Programs



## NCC's LDRD Exploratory Research portfolio



- Quantitative Tomography Simulations and Reconstruction Algorithms—Martz
  - Straight-ray (CTSIM/HADES) and Monte Carlo (COG/MCNP) simulations
  - Validation of simulation models with experimental DR and CT data
  - Better object recovery by incorporation of physics in forward model
- Speckle Reduction for LIDAR—Bowers
  - Speckle compensation through nonlinear-optical techniques
  - Speckle modeling software with nonlinear-optical subroutines
  - Improved performance of LIDAR systems for faster, more accurate, results
- High Resolution and Fast Imaging Photothermal Microscopy—Chinn
  - Imaging of micron-sized defects for 106 improvement in speed
  - Increased imaging sensitivity through optical lock-in hardware
  - Nanometer resolution in scanning using optical fiber tips





## NCC s FY00 Technology Base portfolio



	PI	TITLE
	Priority list of proposals	
1	Landram	Flaw Detection Using Infrared Methods
2	Blaedel **	The conceptual design of a facility for manufacturing meso-scale devices for NIF
3	Chambers	Extended Resolution Diffraction Tomography
4	Decker	Advanced x-ray sources
5	Dolan	Near Nanoscale X-ray Computed Tomography with Low to Medium Energy X-rays
6	Thomas	Ultrasonic Phased Arrays
7	Roberson	True Three Dimensional Signal Processing Tool Development
8	Kallman	Rapid, High Resolution Ultrasound Tomography
9	Jackson	RECON-Code Maintenance

Target Fab./Char. Related \*\* To be presented in PSEC or MMED presentation

hem 4/17/00



NCC-11

## The NCC s projects supports three LLNL programs directly and others indirectly



Focus areas	D&NT	Lasers	NAI	Energy/Env	Bioscience	Other
Quantitative NDE	HR1-Martz T1 Landram, T2-Blaedel; T3-Chambers; T6-Thomas; T7-Kallman; T8 Jackson	H T2-Blaedel; T8 Jackson	M R2-Bowers	H T8 Jackson	H	H
Fast Nanoscale Imaging	H T4-Decker; T5-Dolan	H T3-Chinn T4-Decker; T5-Dolan	M	M T5-Dolan	H	H
Embedded in Design	H	H	M	H	M	H
Multispectral	M	H	H	H	H	M
Portability	M	M	H	H	M	M

### Objectives

**Quantitative NDE** The objective of quantitative multi-dimensional NDE is to obtain measurements to evaluate physical properties: geometry, density, elemental & molecular composition, thermal, structural, mechanical, radioactive assay, etc., in all three dimensions and as a function of time for the materials/objects we characterize. We set a goal of 0.1% accuracy for each of these properties at videoframe rates. This includes integration of inspection, evaluation and characterization methods to predict material response and life times, and when necessary, reduce data to a quantitative decision.

**Fast Nanoscale Imaging** The fast Nanoscale imaging objective is to provide multiple programs with quantitative multidimensional (>3D) imaging down to nanoscale lengths. Most techniques that provide such a capability to date are surface and near surface methods. Our goal is to do this on a specimen in bulk/3D, i.e., beyond nearsurface. We want to develop NDE techniques that can obtain spatial resolution of 100 nm (resolve two objects separated by 100 nm) over 1-mm field-of-view at video frame rates. We need to determine if superresolution is possible and useful to obtain spatial resolutions to better than the diffraction limit of the NDE imaging systems.

<sup>†</sup>D&NT is Defense & Nuclear Technologies, NAI is Nonproliferation, Arms Control & International Security; Others refers to work outside of DOE, eg., DOT, DOD and private sector; Key: H is high; M is medium; and L is low impact to the program; R is LDRD and T is Tech. Base.

10/28/99

hem 4/17/00



NCC-12

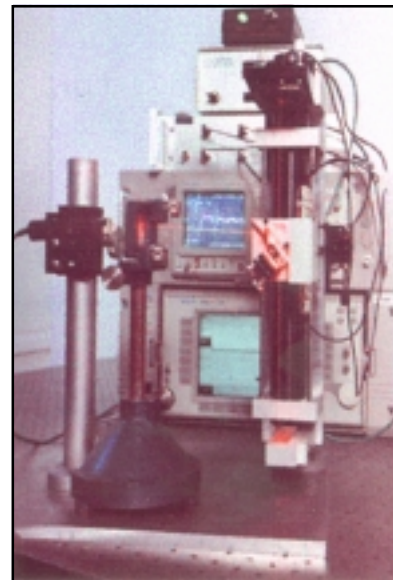
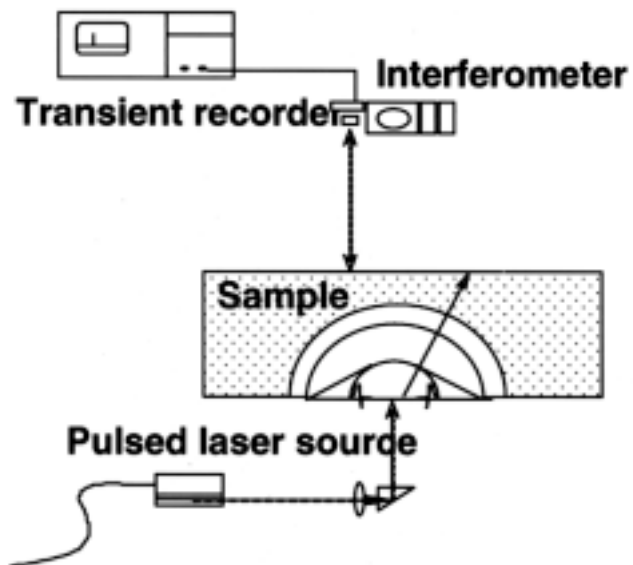


## Ultrasonic Activities at Lawrence Livermore National Laboratory



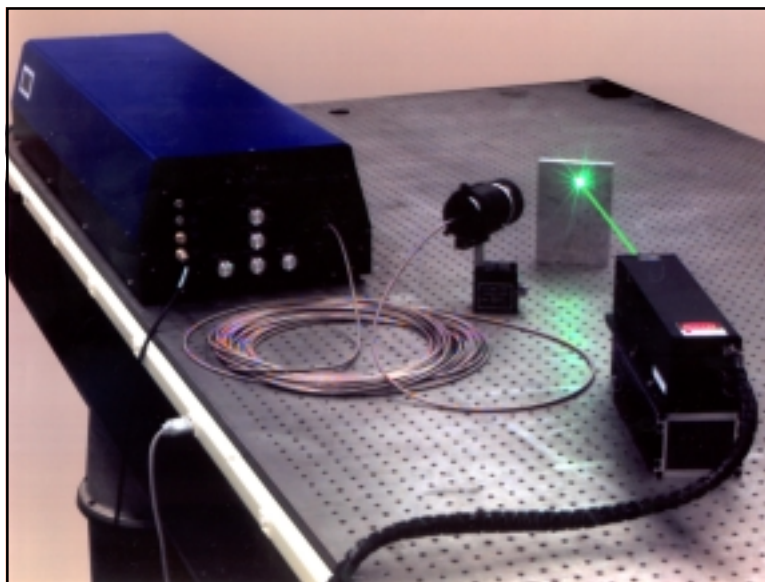
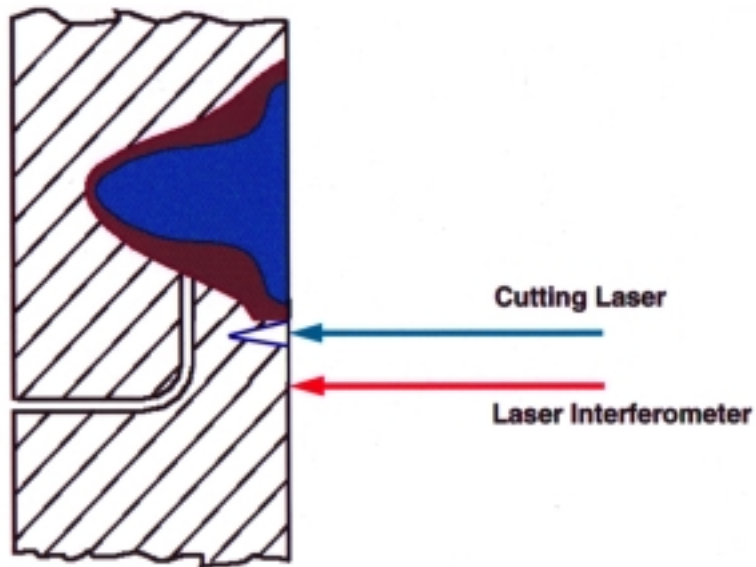
Graham Thomas

Ultrasonic energy can be generated  
and detected with lasers



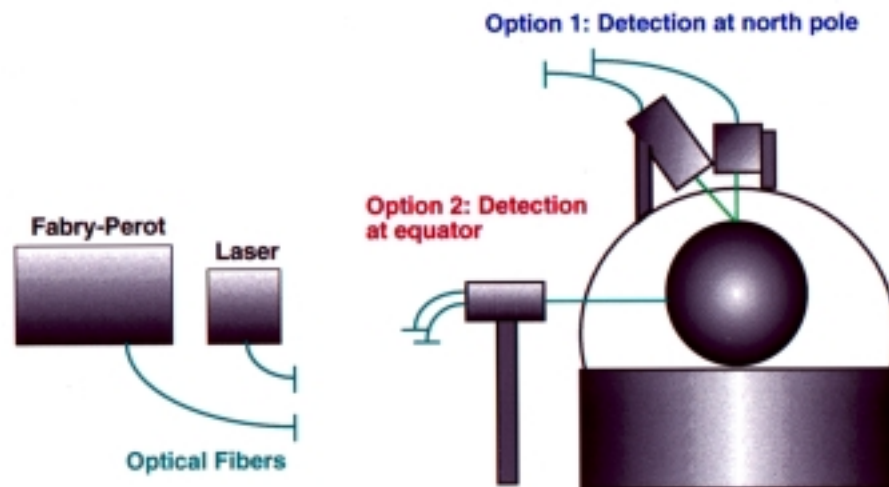


**Laser interferometer will sense the change in the acoustic signal generated by cutting laser**

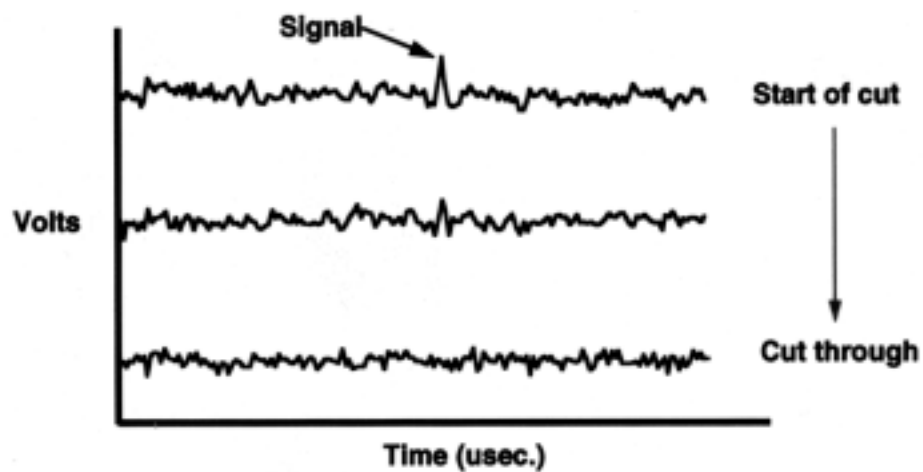




## Ultrasonic detection using Fabry-Perot interferometer



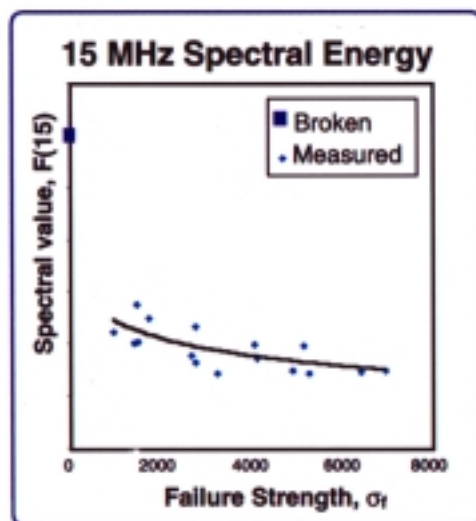
## An ultrasonic signal decreases with cutting laser penetration depth



Data from interferometer pointed at the north pole



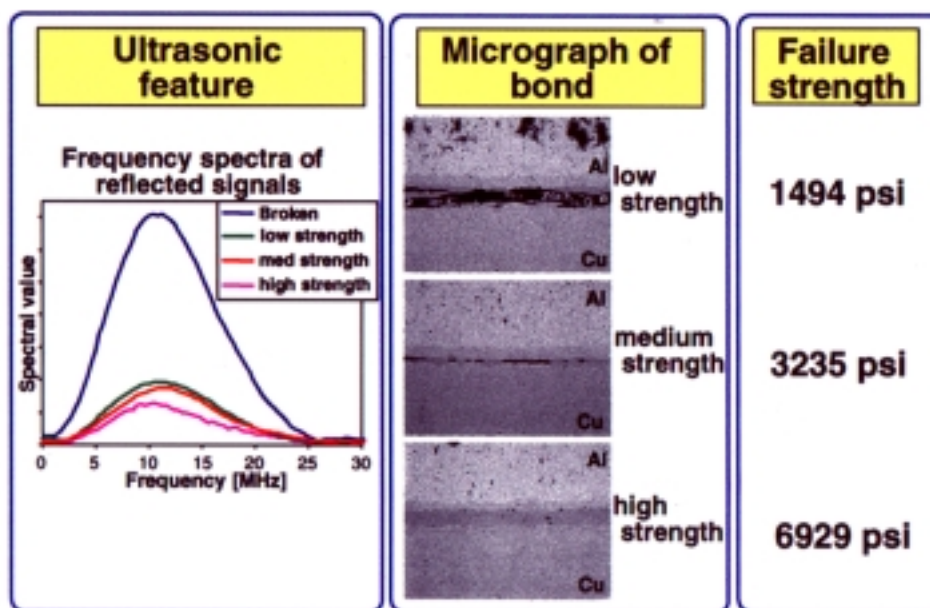
We derived an empirical equation relating failure strength to ultrasonic data



$$\sigma_f = \exp \left[ \frac{F(15) - 287.4}{23.87} \right]$$

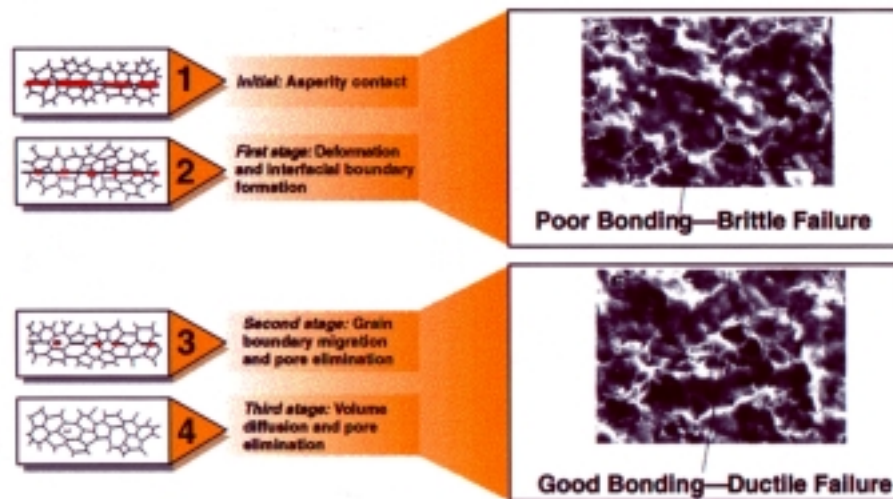
Given a feature value  $F(15)$  we predict failure strength

Ultrasonic features correlate with microstructure and failure strength





## We fabricated bond samples to characterize quality

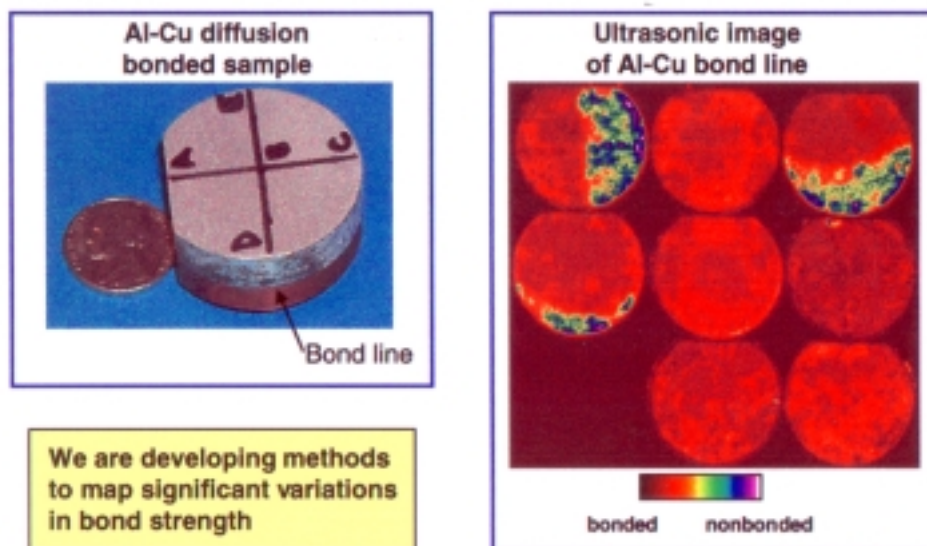


## Tensile specimens were tested to give us the bond quality

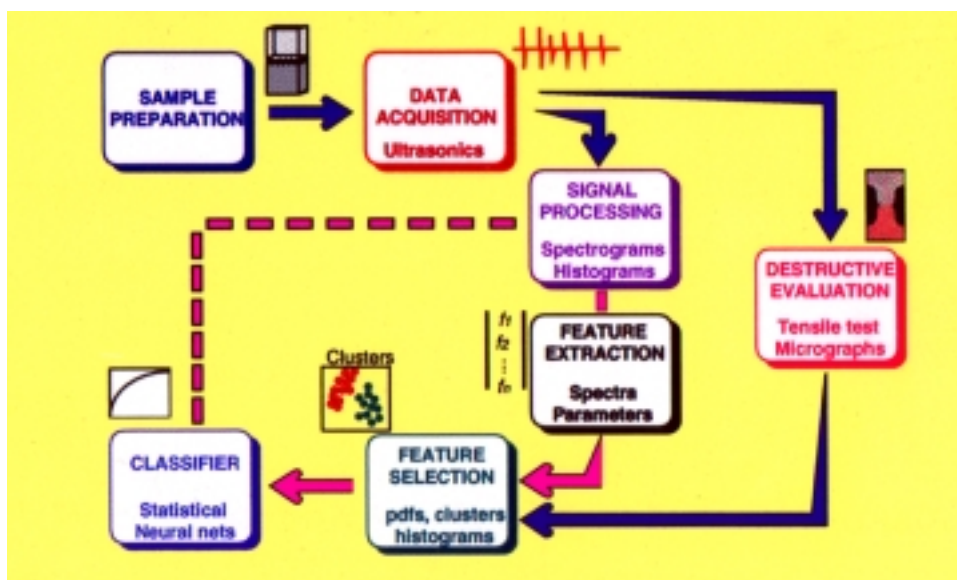




## High resolution images currently display nonbonded regions

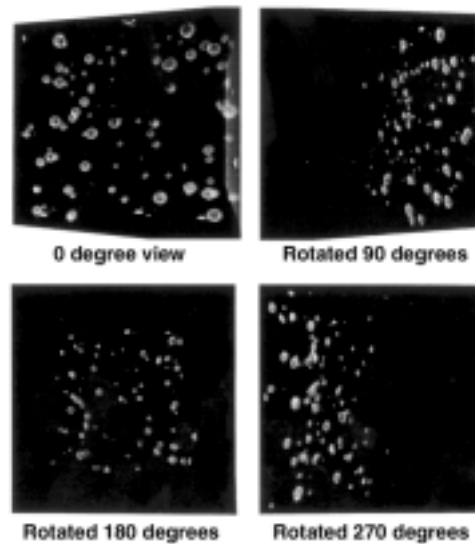


## Experimental protocol has produced a correlation between bond strength and ultrasonic signals

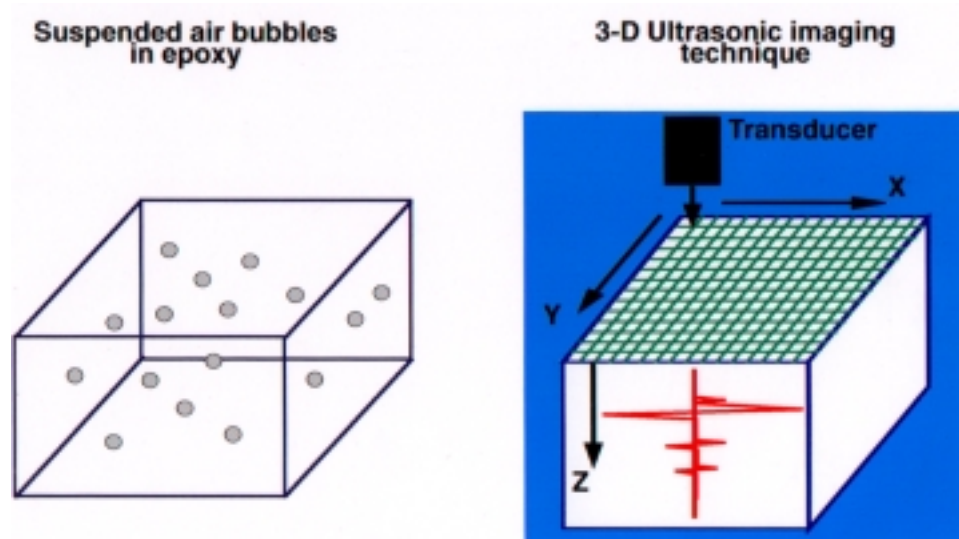




**Ultrasonics displays defects in 3-D so that their size, location, and orientation can be determined**

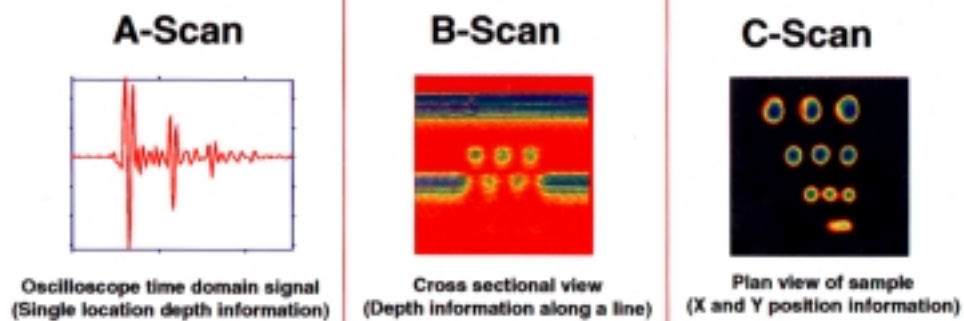


**Combining x and y location with depth information provides 3-D images**

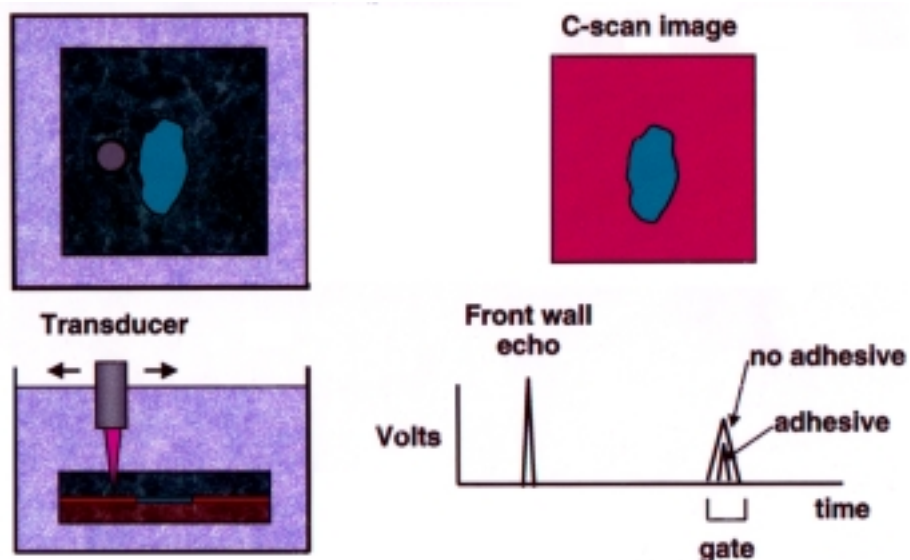




## Ultrasonic data can be presented in several formats

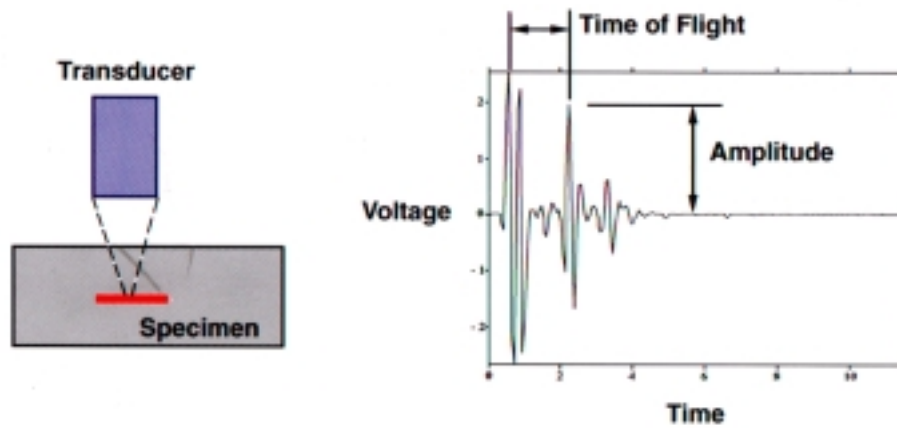


## Ultrasonic C-scans provide planar view of reflectors at a prescribed depth





**Ultrasonic nondestructive evaluation interrogates components with high frequency acoustic energy**









## **Signal Processing for Laser Ultrasonic NDE**

**R. Huber, J. Candy**

LAWRENCE LIVERMORE NATIONAL LABORATORY

### **ORGANIZATION**

---

- **INTRODUCTION: Laser Ultrasonics**
- **SIGNAL PROCESSING: Correlation Cancelling**
- **Summary**

LAWRENCE LIVERMORE NATIONAL LABORATORY



## **LASER ULTRASONICS**

LAWRENCE LIVERMORE NATIONAL LABORATORY

## **Laser-based Ultrasound**

- **Allows non-contact, remote testing.**
- **Allows testing in vessels, furnaces.**
- **Nondestructive.**
- **Linear broadband.**
- **Works with complicated geometries.**

LAWRENCE LIVERMORE NATIONAL LABORATORY

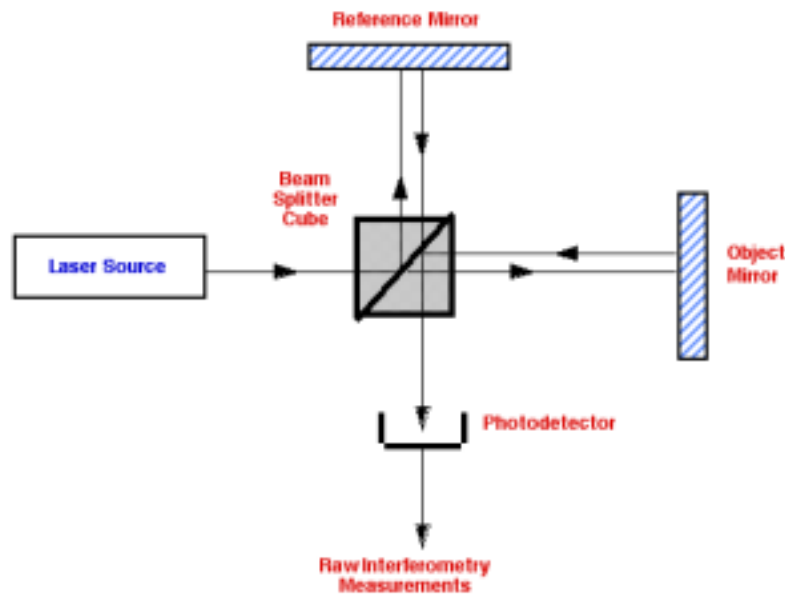


## Laser-Based Ultrasound System

- Pulsed Nd: YAG laser.
- Stabilized HeNe-based Michelson interferometer
- Computer-controlled motion stages.

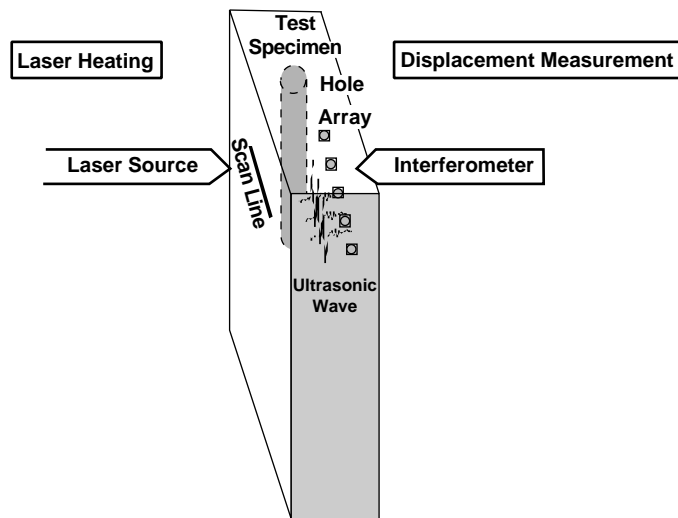
LAWRENCE LIVERMORE NATIONAL LABORATORY

Displacements can accurately be measured using a laser interferometer similar to the simple setup here:



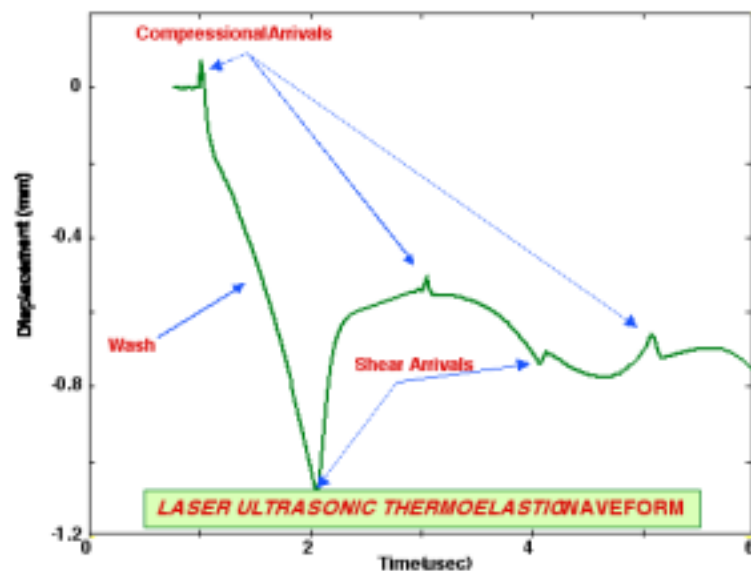


**Ultrasonic waves are generated through the thermoelastic expansion of the material created by laser heating**



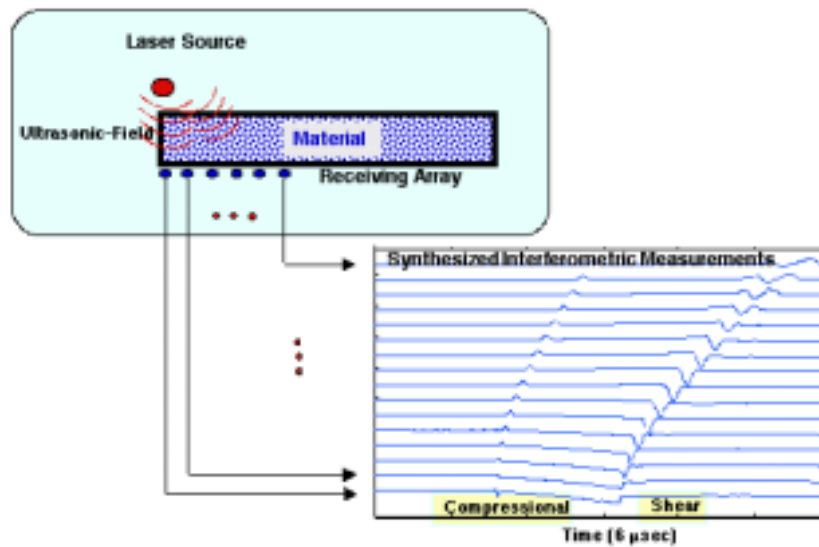
LAWRENCE LIVERMORE NATIONAL LABORATORY

**A typical epicentral waveform has the following characteristics:**





Surface displacements can be measured by a sensitive laser interferometer to create a synthetic multichannel array

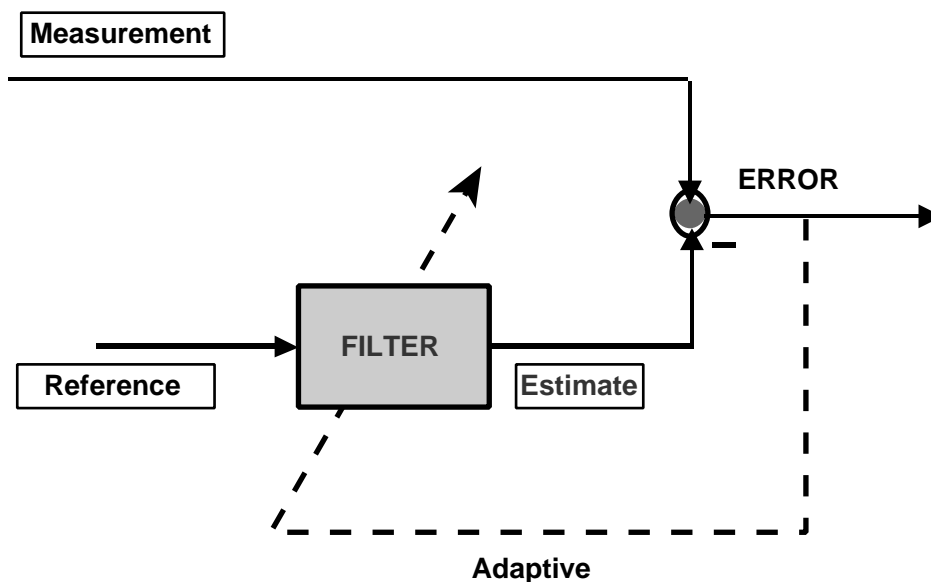


## PRE-PROCESSING

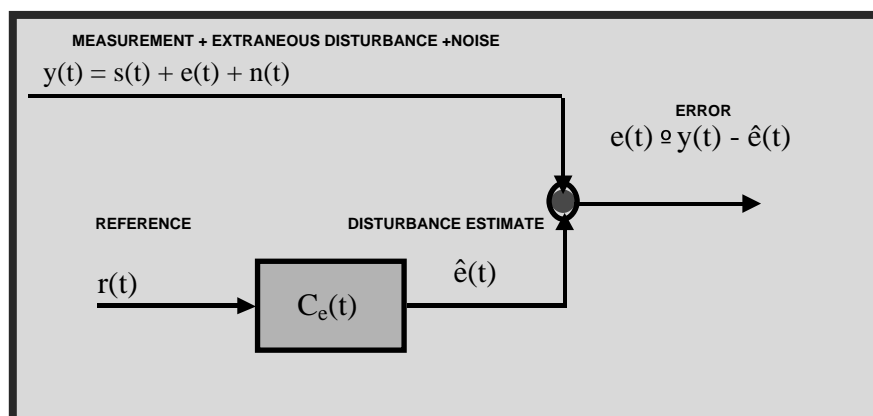
LAWRENCE LIVERMORE NATIONAL LABORATORY



Critical preprocessing is accomplished using a correlation canceller algorithm to remove the “wash and primary arrivals”



### CORRELATION CANCELLER



$$e(t) = y(t) - \hat{e}(t) = [s(t) + e(t) + n(t)] - \hat{e}(t) \gg s(t)$$

$$\hat{e}(t) = C_e(t) * r(t)$$

$$\underline{C}_e = R_{rr}^{-1} \underline{r}_{yr}$$



**CORRELATION/NOISE CANCELLATION:**  
*is a form of optimal mean-squared error estim.*

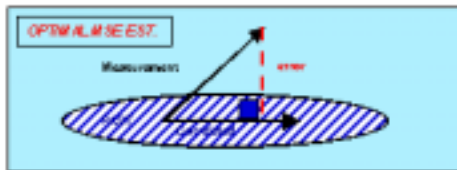
Suppose a measurement can be 'decomposed' into two parts: one **correlated** and one **uncorrelated**

$$y(t) = y_{\text{cor}}(t) + y_{\text{uncor}}(t)$$

where  $y_{\text{cor}}(t) \perp Y(t-1)$  [past data] and  $y_{\text{uncor}}(t) \perp Y(t-1)$  [past data]

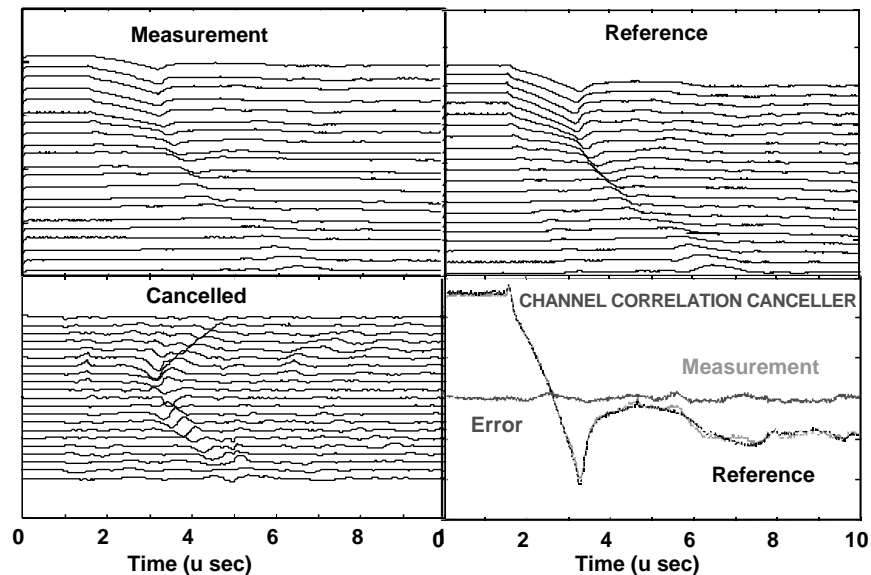
For optimal estimation  $\hat{y}(t) \approx y_{\text{cor}}(t)$  and therefore, we define the 'error' as

$$e(t) \approx y(t) - \hat{y}(t) \approx y_{\text{uncor}}(t)$$



**Correlation cancelling applied to experimental data has proven to be vital to remove the "wash and primary arrivals"**

**Laser Ultrasonic Correlation Cancellation**





## SUMMARY

---

We have developed a **novel correlation canceling** approach to remove unwanted components of signals, and have applied it to actual experimentally obtained laser-based ultrasound signals.

LAWRENCE LIVERMORE NATIONAL LABORATORY



**MATCHED-FIELD IMAGING  
of  
ULTRASOUND  
for  
NONDESTRUCTIVE EVALUATION**

**J. Candy, R. Huber, D. Chambers, G. Thomas**

CASIS Workshop

Livermore, CA  
November 11-12, 1999

**LAWRENCE LIVERMORE NATIONAL LABORATORY**

**ORGANIZATION**

---

- **INTRODUCTION:**
- **MATCHED-FIELD Imaging**
- **Results**
- **Summary**

**LAWRENCE LIVERMORE NATIONAL LABORATORY**



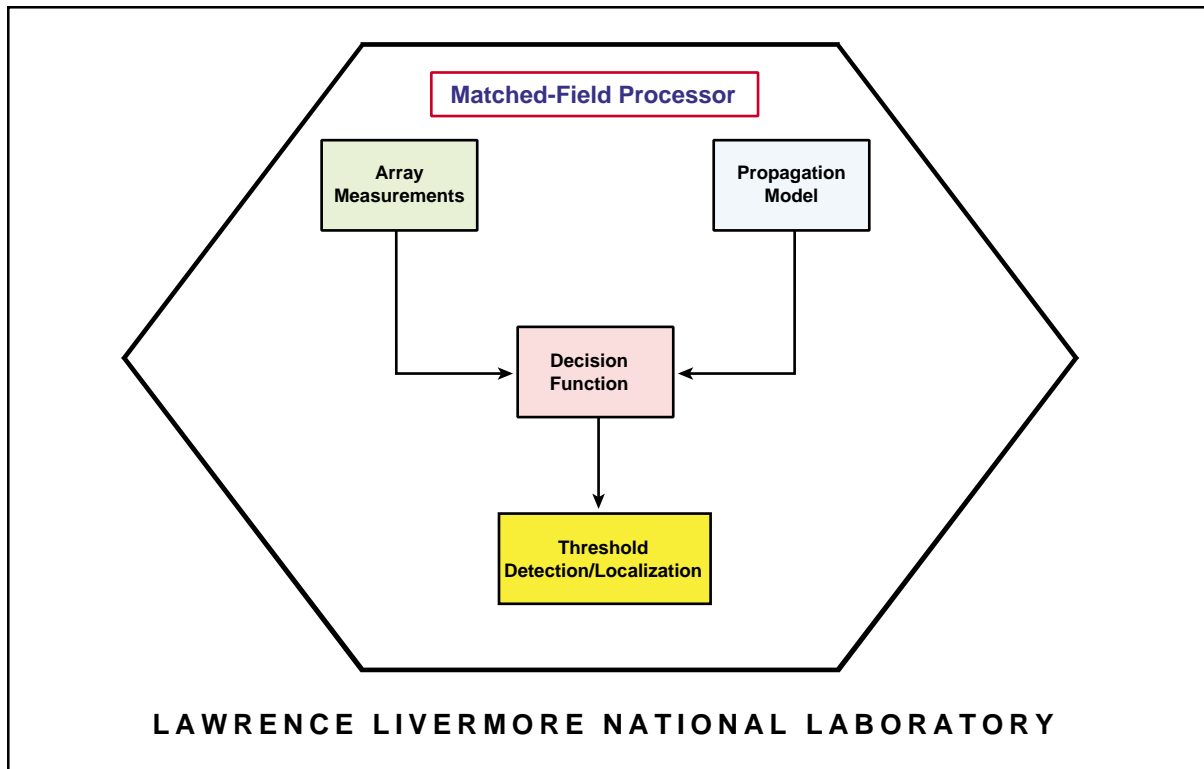
## MATCHED-FIELD IMAGING APPROACH

### MATCHED-FIELD IMAGING is:

- Is a technique to **detect**, **localize** and **image** flaws for NDE
- Is the **optimal** solution to a basic “**flaw detection**” problem where a known (or modeled) field is used
- Is the “**matched-filter**” solution to a known signal in noise
- The extension of **matched-field processing** (ocean acoustics) to the ultrasonic imaging problem
- Is a **generalization** of the Synthetic Aperture Focusing Technique (SAFT) approach to imaging

LAWRENCE LIVERMORE NATIONAL LABORATORY





**When the position of the “flaw” is unknown, we must perform a composite hypothesis test to solve the detection problem, i.e. ,**

The **composite hypothesis** test is:

$$\begin{aligned} H_0 : \underline{d}(t) &= \underline{n}(t) && [\text{noise only}] \\ H_1 : \underline{d}(t) &= \underline{m}(t; \theta) + \underline{n}(t). && [\text{signal + noise}] \end{aligned}$$

which leads to a **generalized likelihood ratio test** which can be summarized by:

$$\begin{aligned} \max_{\theta_s} P(\underline{\theta}_s) &\begin{matrix} H_1 \\ > \\ < \\ H_0 \end{matrix} \lambda_{\theta} \\ \text{where} \\ P(\underline{\theta}_s) &= \frac{\langle \underline{m}'(T; \underline{\theta}_s) * \underline{d}(T)^2 \rangle}{\langle \underline{m}'(T; \theta_s) * m(TT; \theta_s) \rangle} \end{aligned}$$

LAWRENCE LIVERMORE NATIONAL LABORATORY



### The signal model that we developed is the geometric optics simplification evolving from the elastic wave equation

The model has both **compressional** and **shear** components

$$\underline{f}(t) = \underline{G}_c(t) * s_c(\underline{r}, t) + \underline{G}_s(t) * s_s(\underline{r}, t),$$

which can be simplified significantly in a **homogeneous** medium to:

$$\underline{m}(T) = \underline{f}(T - t) = \underline{G}_c(T - t) + \underline{G}_s(T - t)$$

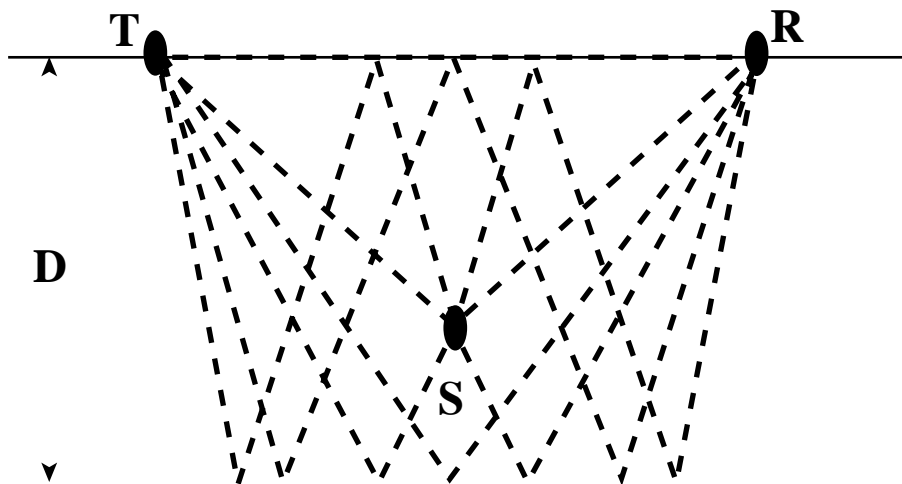
#### NOTE:

- This is the **time reversed, matched-filter** solution
- **Ignoring** either the compressional or shear velocity, we obtain the **SAFT** solution

\*\*\*\* A CONVENTIONAL SUM-DELAY BEAMFORMER \*\*\*\*

LAWRENCE LIVERMORE NATIONAL LABORATORY

### The “matching function” is generated from a propagation model which we generate from the geometric optics viewpoint

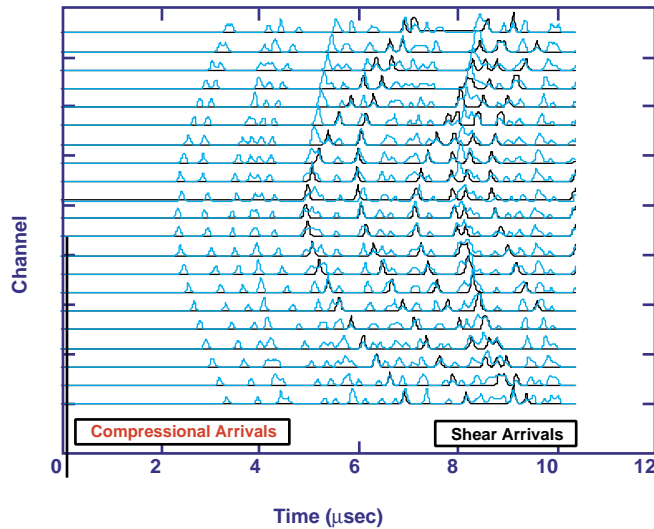


LAWRENCE LIVERMORE NATIONAL LABORATORY



**A typical geometric optics simulation of an aluminum part of 9.5mm in width and 20mm in length is given by:**

---



**LAWRENCE LIVERMORE NATIONAL LABORATORY**

**The Optimal Matched-Field Imaging ALGORITHM is:**

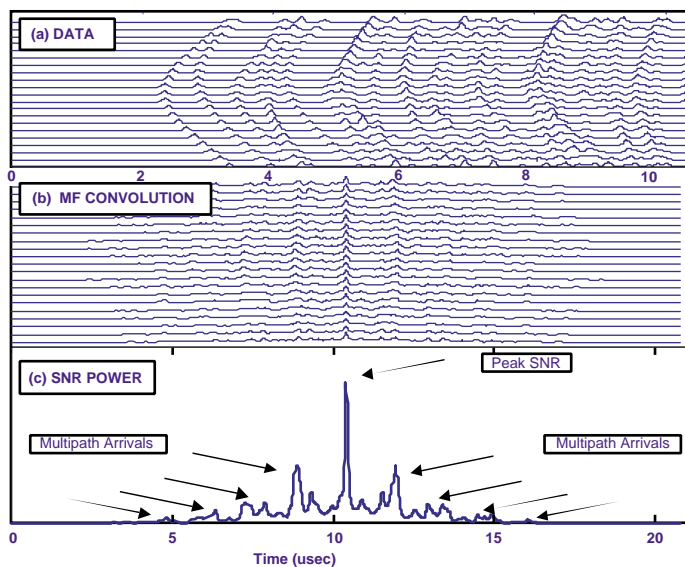
---

- Decompose the test specimen into rectangular pixels specified by the desired spatial sampling intervals, ( $D_x$ ,  $D_y$ )
- For each pixel generate the model replicant by backpropagating the field to the array at each sensor position using the propagation model
- Perform the required convolutions, squaring and summing operations
- Threshold filter the image so that the maximum threshold is satisfied
- Search for maxima to “detect” and “localize” flaws

**LAWRENCE LIVERMORE NATIONAL LABORATORY**

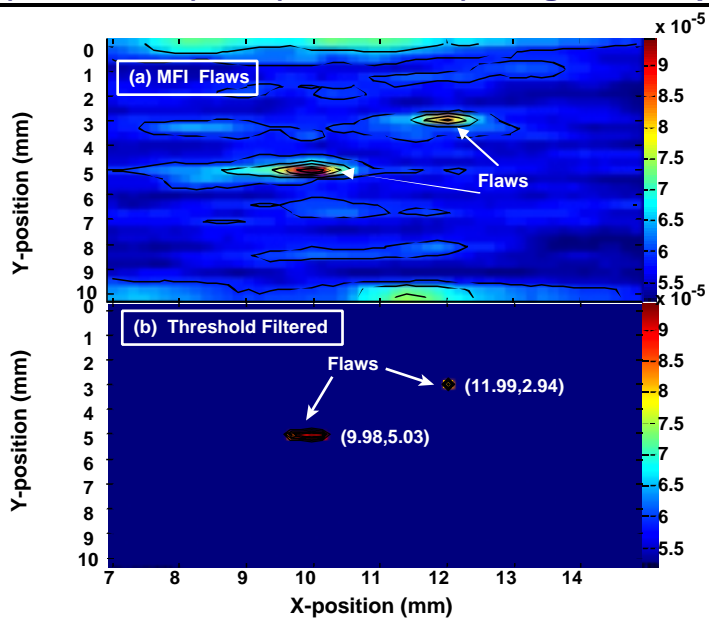


Applying the “matching function” generated from the geometric optics propagation model shows the optimal solution at the flaw



LAWRENCE LIVERMORE NATIONAL LABORATORY

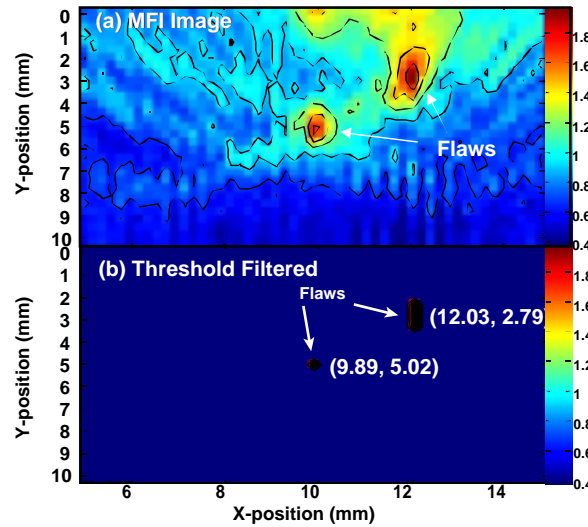
The “optimal” MFI processor works well on simulated data with flaws at (10mm, 5mm) and (12mm, 3mm) using the multipath model.



LAWRENCE LIVERMORE NATIONAL LABORATORY



The “**SUBoptimal**” SAFT processor performs reasonably on the simulated data with flaws at (10mm, 5mm) and (12mm, 3mm) using a homogeneous model without multipath.

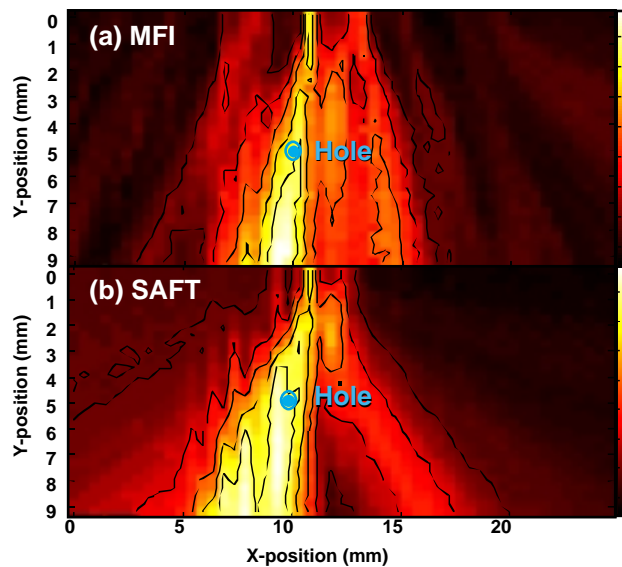


LAWRENCE LIVERMORE NATIONAL LABORATORY

## RESULTS

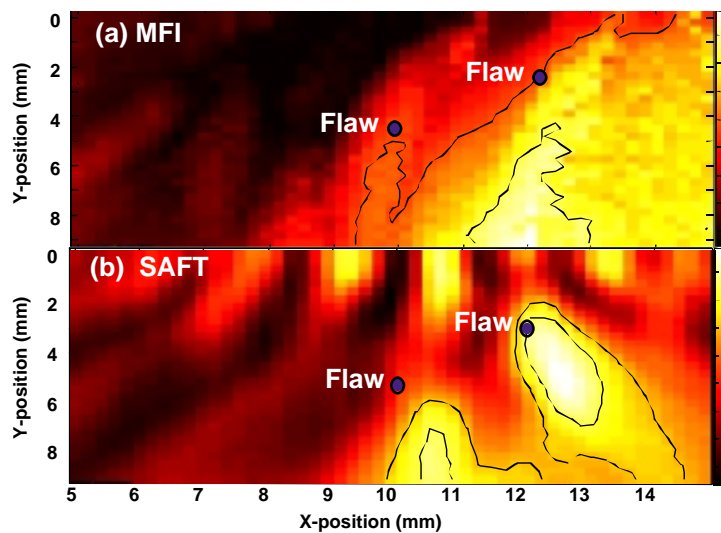


**Epicentral scans of an aluminum part of 9.5mm in width and 20mm in length with a hole at (10mm,5mm) provide a flaw detection**



**LAWRENCE LIVERMORE NATIONAL LABORATORY**

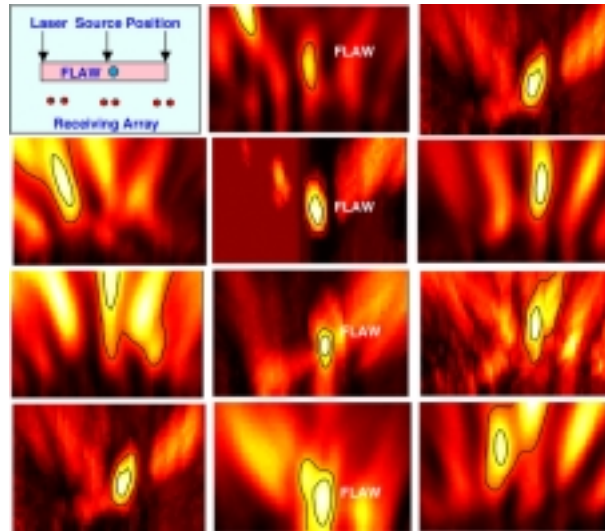
**Epicentral scans of an aluminum part of 9.5mm in width and 20mm in length also detect flaws (holes) at (10mm, 5mm) and (12mm, 3mm)**



**LAWRENCE LIVERMORE NATIONAL LABORATORY**



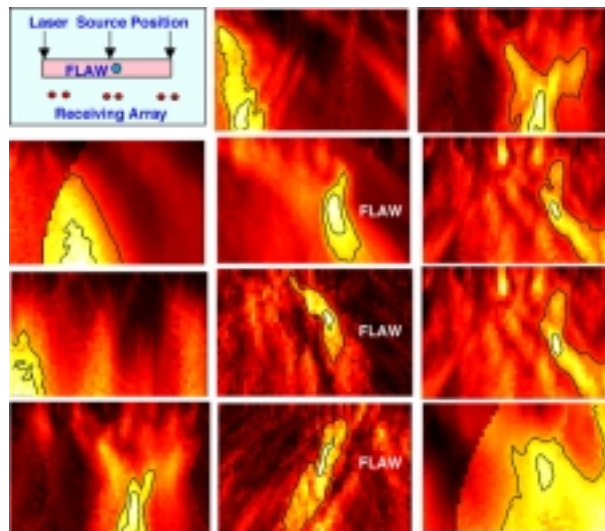
A decoupage of images, one for each source location, provides an evolution of the flaw detection and **MFI Shear Only process**



MFI LASER ULTRASONIC IMAGING (Shear)

LAWRENCE LIVERMORE NATIONAL LABORATORY

A decoupage of images, one for each source location, provides an evolution of the flaw detection and **MFI homogenous process**



LAWRENCE LIVERMORE NATIONAL LABORATORY



# SUMMARY

---

We have developed a “matched-field imaging” approach to imaging laser ultrasonic data obtaining by creating a synthetic spatio-temporal array matched filter processing using a multipath, homogeneous medium model demonstrating the feasibility of such an approach on both simulated and experimental data

LAWRENCE LIVERMORE NATIONAL LABORATORY



## Void Analysis from CT Imagery with Applications to Damage Evolution in an AM60B Magnesium Alloy

*Amy M. Waters, Harry E. Martz, Kenneth W. Dolan,  
Mark F. Horstemeyer<sup>2</sup>, Earl O. Updike, and Robert E. Green, Jr.<sup>3</sup>*

Lawrence Livermore National Laboratory (LLNL) and Sandia National Laboratory (SNL) are collaborating on the development of new techniques to study damage evolution and growth in material specimens subjected to mechanical testing. These techniques include metallography, radiography, computed tomography (CT) and modeling. The material specimens being studied include cast magnesium and aluminum alloys, and forged stainless steel. We will concentrate on characterizing mechanically deformed magnesium alloy specimens using computed tomography. Several notched tensile magnesium specimens were uniaxially loaded to obtain different levels of deformation. Specimens were initially characterized by radiography and computed tomography to determine the preloaded state. Subsequent CT scans were performed at different displacement percentages. The CT volumetric data are being used to measure void volume fraction in all three dimensions nondestructively to determine the effect of void growth on the mechanical behavior of the materials.

*2 Permanent address Sandia National Laboratory, Livermore, CA 94550*

*3 Permanent address Center for Nondestructive Evaluation, 102 Maryland Hall, The Johns Hopkins University, Baltimore MD 21218*



# Void Analysis from CT Imagery with Applications to Damage Evolution

prepared by

Amy M. Waters, Harry E. Martz,  
Ken Dolan, and Derrill Rikard  
Nondestructive Evaluation Section  
Lawrence Livermore National Laboratory

Mark Horstemeyer  
Sandia National Laboratories, California

Robert E. Green, Jr.  
The Johns Hopkins University



presented at

CASIS Workshop  
LLNL

November 11 - 12, 1999  
Livermore, California

## Agenda



- Background of project
- Overview of thesis research
- Magnesium alloy tensile bars
- Mechanical loading
- CT simulation studies
- CT data acquisition
- CT data processing
- Preliminary results
- Void analysis
- Summary and future work



amw 4/17/00

1



## **We are using CT to help develop cast light metals for automotive components (Sponsor: USAMP)**



- **Sponsor: United States Automotive Materials Partnership (USAMP), Lawrence Livermore National Laboratory, and the Center for Nondestructive Evaluation, Johns Hopkins University**
- **Design and Product Optimization for Cast Light Metals**
  - **Microstructure Characterization - Arun Gokhale, Georgia Institute of Technology**
  - **Solidification Model - S. Viswanathan, Oak Ridge National Laboratory**
  - **Database - Jim Sudy, Westmoreland Mechanical Testing and Research**
  - **Validation - All team members**
  - **Crack Nucleation - Jinghong Fan and David McDowell, Georgia Tech**
  - **Property Model - Mark Horstemeyer, Sandia National Laboratory**
  - **Design Guide - Greg Sanders, Entelechy**
  - **Monitoring/Quality Assurance - Ken Dolan, LLNL**

amw 4/17/00

2

## **Microstructure Characterization: goals/objectives**



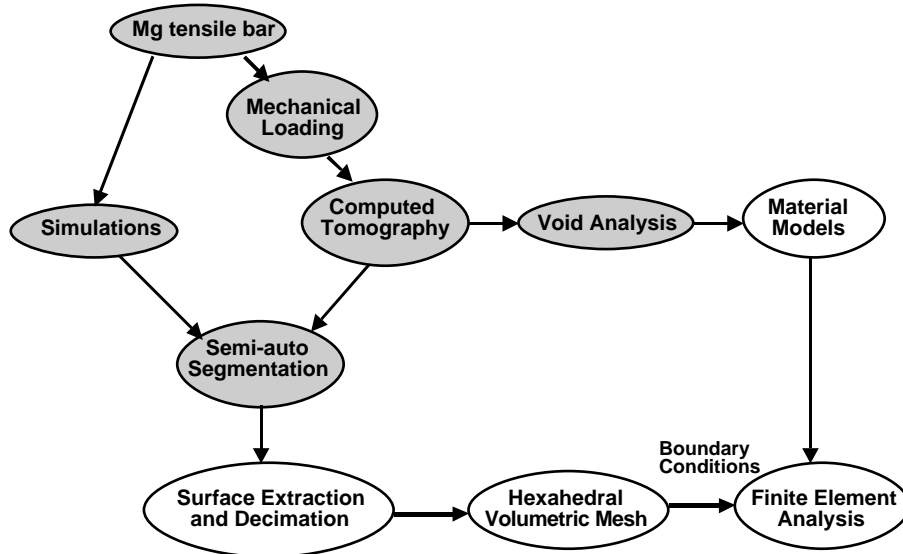
- **Input initial microstructural state (porosity, etc.) of castings into material model and perform structural analysis (FEM) on castings to simulate in-service performance**
  - **Characterize behavior of Mg-alloys**
    - **Characterize internal structure, e.g., defects, voids, inclusions, cracks**
    - **Create material models to predict behavior based on internal structure perform finite element modeling based on material models and structure**
    - **Validate models**
    - **Mechanical testing**

amw 4/17/00

3



## Summary of the steps required to characterize and eventually predict the behavior of new materials



amw 4/17/00

4

## We are studying 9 magnesium alloy tensile specimens



- Mg-alloy is AM60B containing 6% aluminum and 0.377% manganese by weight
- Samples were cast using a cold chamber die casting machine with a 600-ton locking force, injection temperature of 675 to 690 celsius
- Three different notch geometries, Series F, G, H

Series	Max. Outer Diameter	Notch Radius
F	1.265-cm	1.27-cm
G	1.265-cm	0.794-cm
H	1.265-cm	0.635-cm



H

G

F

amw 4/17/00

5



**We are using CT throughout the mechanical loading of the tensile bars to better understand the dynamics of damage evolution**



- Samples will be loaded to percentages of the failure load
  - 60%, 87%, 93% and 100%
- Failure loads were determined by Westmoreland Mechanical Testing, PA

Series	Notch Radius (cm)	$\sigma_{UTS}$ (MPa)	60% load (N)	87% load (N)	93% load (N)
H	0.635	207	9176	13304	14225
G	0.794	221	9283	13460	14389
F	1.27	241	9879	14327	15314

amw 4/17/00

6

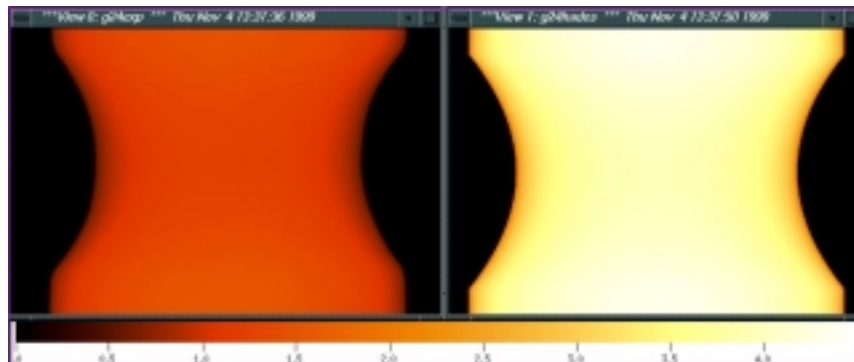
**We are using a radiographic simulation code (HADES) to optimize CT acquisition, data processing and analysis**



HADES is a code developed at LLNL to simulate radiographs. We are using HADES to model Mg-alloy tensile specimens with KCAT scanner geometry and parameters.

**Experimental radiograph  
(75 KV w/0.08 mm Al filter)**

**Simulated radiograph  
(75 KV w/0.08 mm Al filter)**

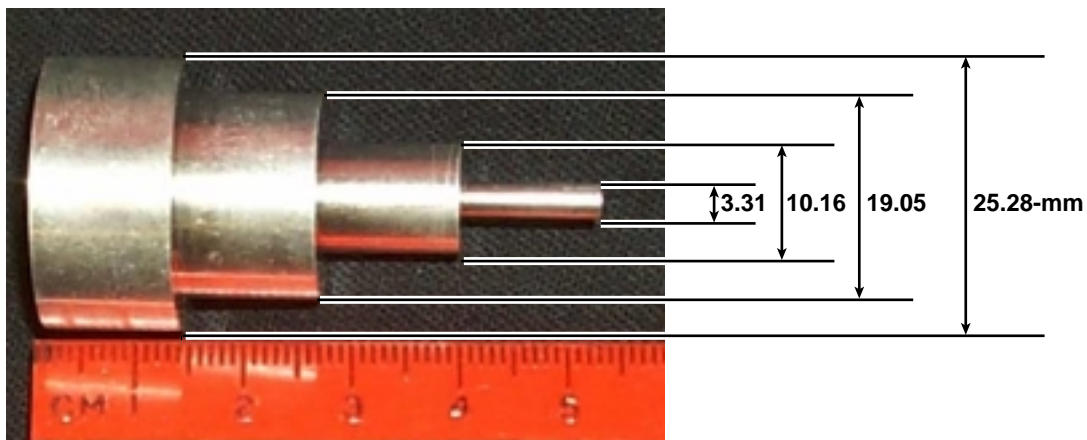


amw 4/17/00

7



**We have created a phantom made of pure magnesium (99.9%) to study artifact reduction and to help optimize CT data acquisition**



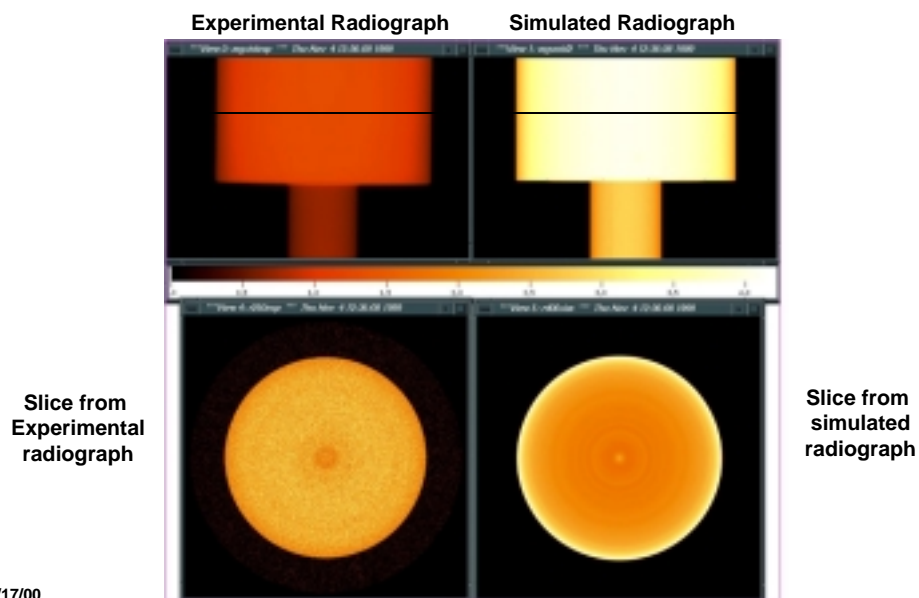
amw 4/17/00

8

**We have simulated the pure Magnesium Phantom**



**We will include voids and various objects in the simulated phantom to determine effect on CT reconstructions**

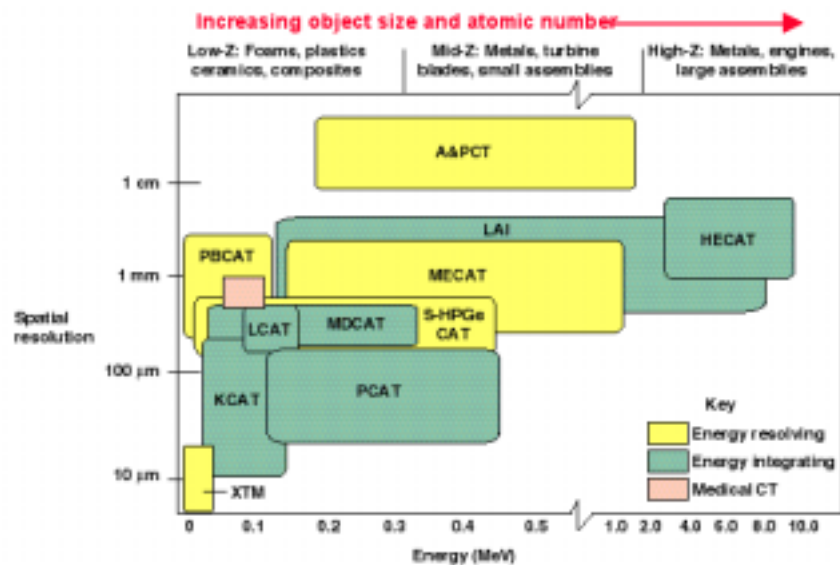


amw 4/17/00

9



We have researched, developed and applied several CT scanners including data processing and image reconstruction and analysis software



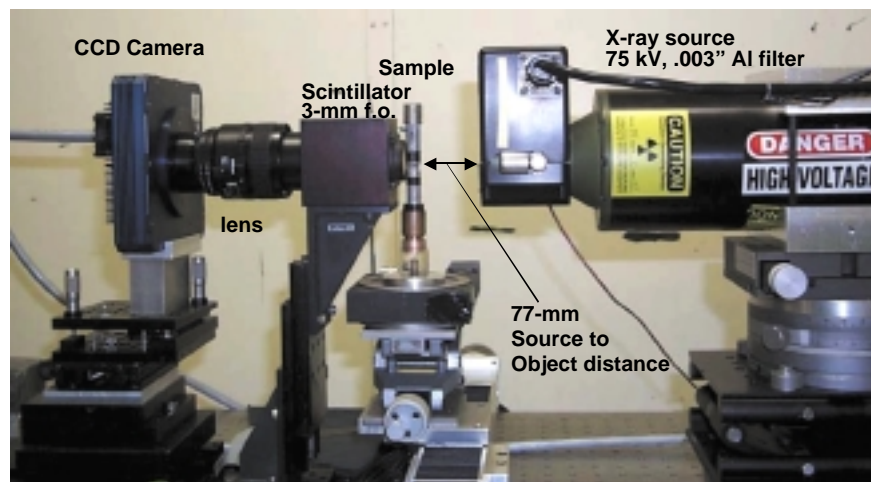
amw 4/17/00

We have used PCAT and KCAT to study the magnesium tensile bars

10

## Experimental Setup - KCAT

KCAT is a high-spatial resolution CT scanner



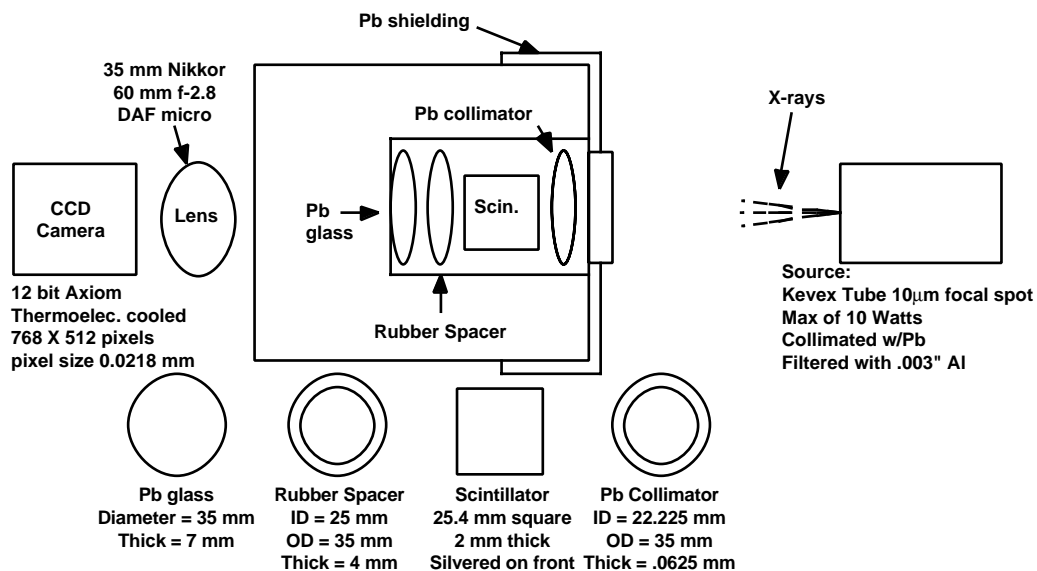
We acquired 360 projections over 360 degrees. Acquisition time was approximately 5 hours. Modulation of the system is approximately 0.5 at 10 lp/mm.

amw 4/17/00

11



## Schematic of scintillator/lens/CCD detector that we are using to study the magnesium tensile bars

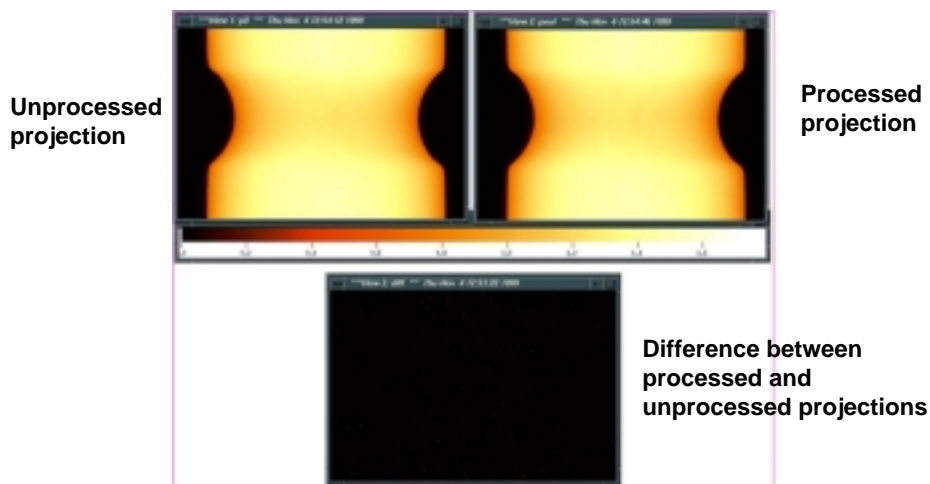


amw 4/17/00

12

## Outlier Removal

- An outlier removal routine is applied to all projections [ $\ln(I_0/I)$ ] before processing
  - A median filtered image is created and compared to original image. Any pixels with values outside a given difference are replaced by the median filtered value.



amw 4/17/00

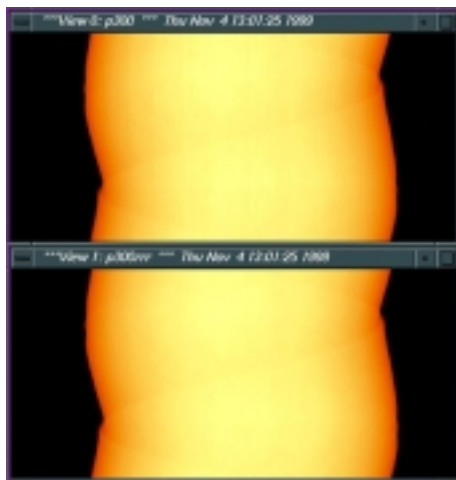
13



## Ring Removal

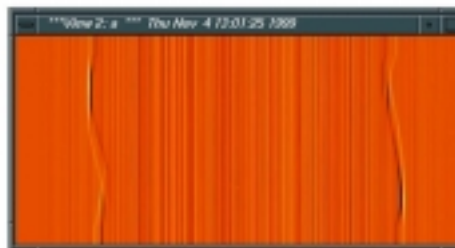


Rings are due to an imbalance between detector elements. Sinograms were created and processed to removed ring artifacts. Rings appear as vertical lines in the sinogram and are minimized by using a combination of smoothing filters.



Unprocessed sinogram (with rings)

Difference



Processed sinogram (without rings)

amw 4/17/00

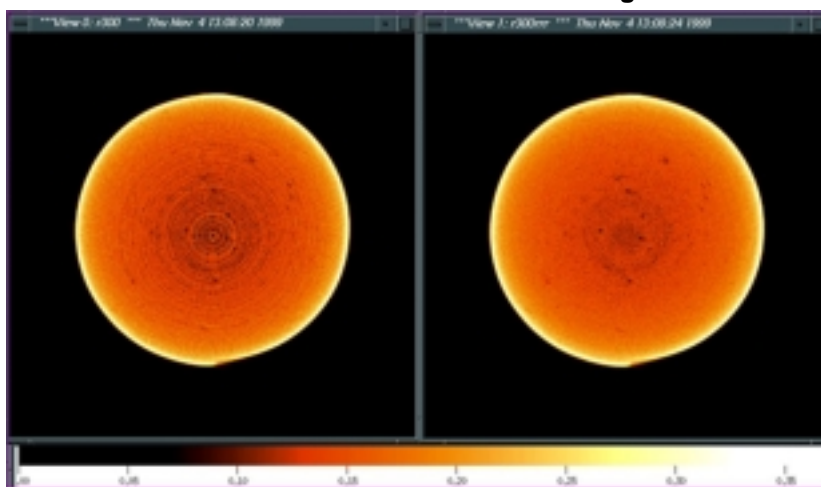
14

## Ring removal helps simplify data interpretation and analysis



Without ring removal

With ring removal

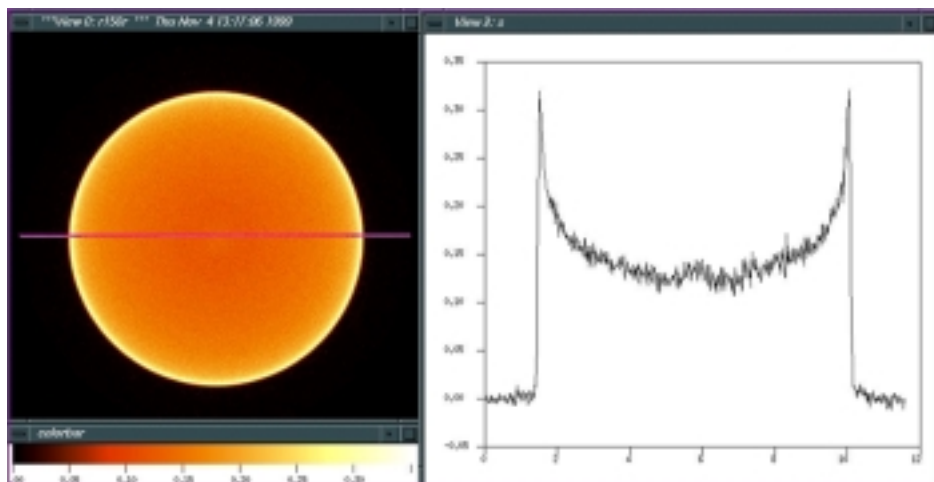


amw 4/17/00

15



**Beam hardening is a result of the preferential absorption of low energy x-rays, resulting in a “cupping” artifact**

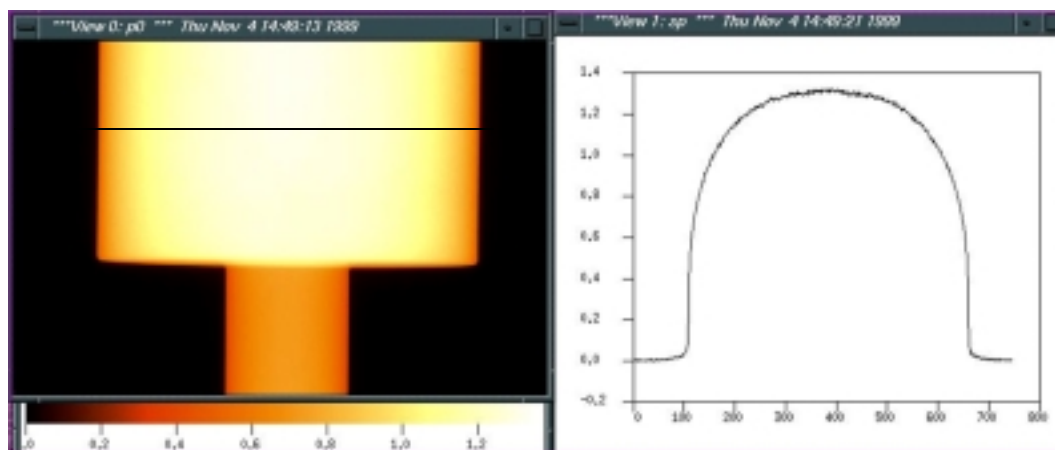


One reconstructed CT slice without any beam hardening corrections, and a 1-D profile. Note the apparent density gradient from outside to inside of part.

amw 4/17/00

16

## Beam Hardening



Radiograph of two smallest sections of Mg beam hardening phantom, 10.16 and 3.31-mm outer diameters.

One-dimensional profile from the larger section of Mg phantom. This data is used to obtain the function between chord length and attenuation times length ( $\mu \cdot l$ )

amw 4/17/00

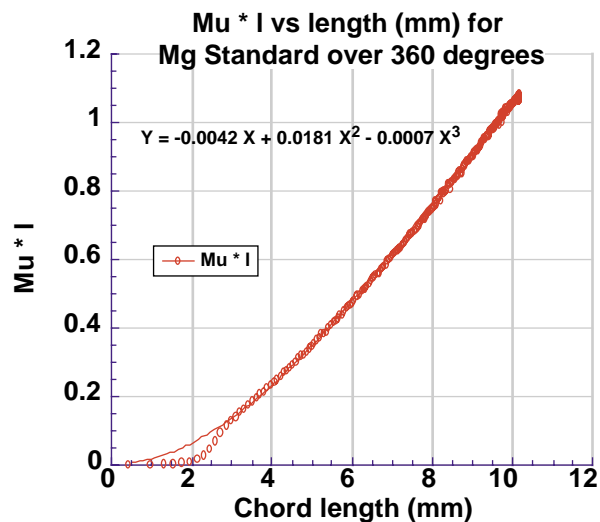
17



## Beam Hardening Correction



Plot of attenuation vs. chord length. The best fit polynomial is used on the original sinogram to correct for beam hardening.



amw 4/17/00

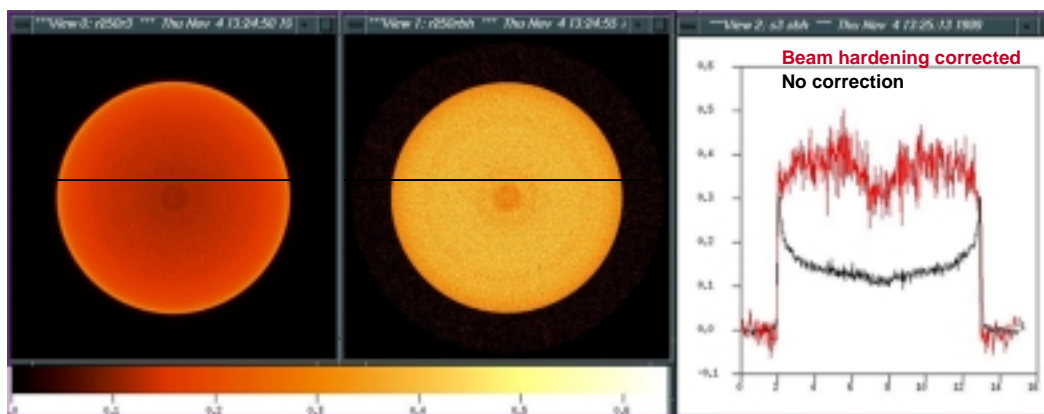
18

## Beam Hardening Correction



No beam hardening correction

With beam hardening correction



Note: the beam hardening correction increases noise and raises value of CT image.

amw 4/17/00

19



## We are studying different quantitative void analysis methods to reduce the CT data

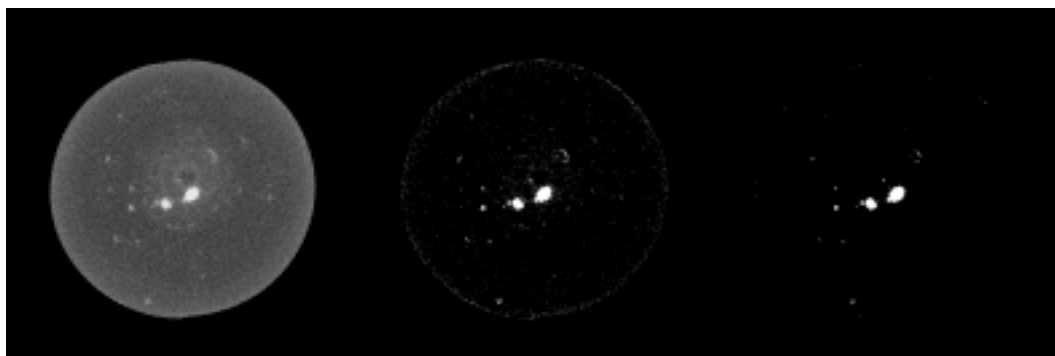


- We are determining void size, orientation and distribution in 3-dimensions using codes developed at LLNL
- Using a 'Cluster Labeling' routine (Kinney)
  - Create an inverted binary 3-D image (void=1)
    - Invert masked image
    - Threshold by intensity value determined from histogram
    - Apply a median filter of size 3 to thresholded image
  - Look for cluster size and connectivity:
    - Number/Volume
    - Volume fraction
    - Size and frequency distribution
    - Nearest neighbor distance

amw 4/17/00

20

## Preliminary cluster analysis results - applied to a single 2-D slice



Masked, inverted image  
(voids appear light)

Thresholded image  
(determined from histogram)

Median filtered image  
(3x3x3)

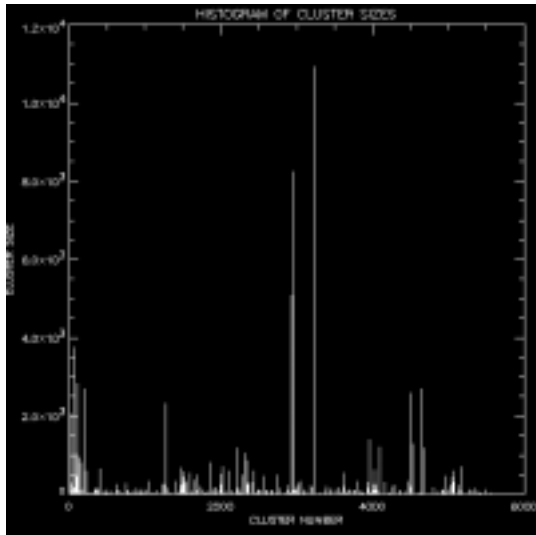
Once the processing parameters are determined the entire CT data set is processed to obtain the 3-D void structure. This data is being incorporated into constitutive material models.

amw 4/17/00

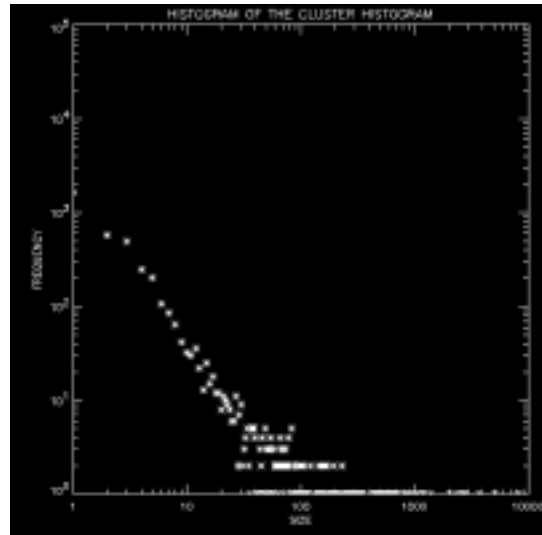
21



## Cluster analysis results



Number and size of voids



Frequency distribution of voids

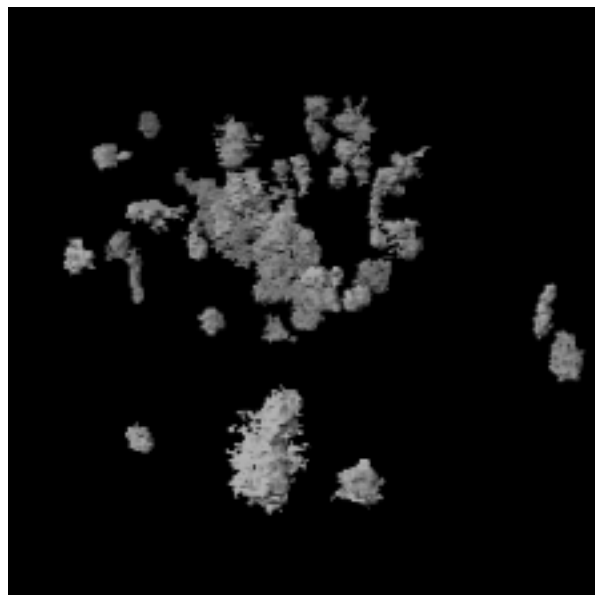
amw 4/17/00

22

## Cluster analysis results



Example of a 3-D surface rendered image of selected voids. This data can be surface meshed and then used with a finite element code for further analysis.



amw 4/17/00

23



## Summary



- Simulation code (HADES) is being used to simulate radiographs of tensile specimens to help optimize CT data acquisition, processing and analysis
- CT data have been acquired for all nine samples before loading
- All samples have been loaded to 60% of their failure loads and rescanned
- Began quantitative void analysis for three Mg tensile bars (one of each notch geometry)



amw 4/17/00

24

## Future Work



- Mechanically load samples to remaining percentages of failure; 87%, 93%, and 100%, and rescan
- Reconstruct all CT data from experimental and simulated radiographs
- Quantitatively characterize voids e.g., size, orientation and 3-D distribution for all tensile bars, before and after loading
- Segment, decimate and mesh both experimental and simulated data, and perform finite element analysis
- Use empirical void analysis results to change parameters in the constitutive model equations to better fit damage evolution results

amw 4/17/00

25



---

---

## **Ultrasonic Techniques for Materials Characterization and Process Monitoring: Evaluation of Low Hydrogen Embrittlement Platings on High Tensile Strength Steel**

**L. P. Martin and M. Rosen**  
Johns Hopkins University

and

**E.A. Lindgren**  
Industrial Quality, Inc.

**Sacrificial plating is currently applied to high  
tensile steels used in landing gear applications**

---



- **Plating is often based on the Ti-Cd system**
- **Typical substrate materials are 4330M and 4340M steel**
- **Significant levels of residual hydrogen may be present after plating**
- **The residual hydrogen is removed by a subsequent heat treatment (hydrogen bakeout)**



### **The plating process is designed to insure that the electroplated layer remains porous**



- The hydrogen removal is critical since these steels can be subject to hydrogen embrittlement cracking
- In order for the bakeout process to be effective, however, it is crucial that the residual hydrogen be able to escape through the plating
- The electroplating procedure is designed to provide porous plating
- Plating characteristics can be affected by bath contamination, current densities, and other process parameters

lpm 4/17/00

NDE-2

### **On line monitoring the porosity level of the LHE plating is necessary in order to eliminate waste**



- Improper plating can lead to hydrogen embrittlement of the component
- Direct testing of individual parts is desirable in order to expedite the manufacturing process and to insure product quality
- Direct testing of individual parts is desirable in order to expedite the manufacturing process and to insure product quality
- Currently used techniques rely on surrogate testing

lpm 4/17/00

NDE-3



## **A surface acoustic wave technique is being evaluated for on line characterization of the LHE plating**

---



- **The technique is nondestructive and nonintrusive**
- **The measurements can be performed rapidly in real-time, and can be used to evaluate every plated component**
- **Surface waves travel along complex shapes such as curves and corners, so that a wide range of sample geometries can be evaluated**

lpm 4/17/00

NDE-4

## **Experimental methodology**

---



- **Measurements of the velocity and amplitude of 5 MHz surface Rayleigh waves have been performed upon LHE plated samples**
- **A broadband voltage spike was used to excite a commercial 5 MHz longitudinal wave transducer**
- **This transducer was coupled, using a commercial gel couplant, to an aluminum mode conversion block**
- **The conversion block was dry coupled to the sample surface with a thin layer of latex**

lpm 4/17/00

NDE-5



## Experimental methodology (cont.)

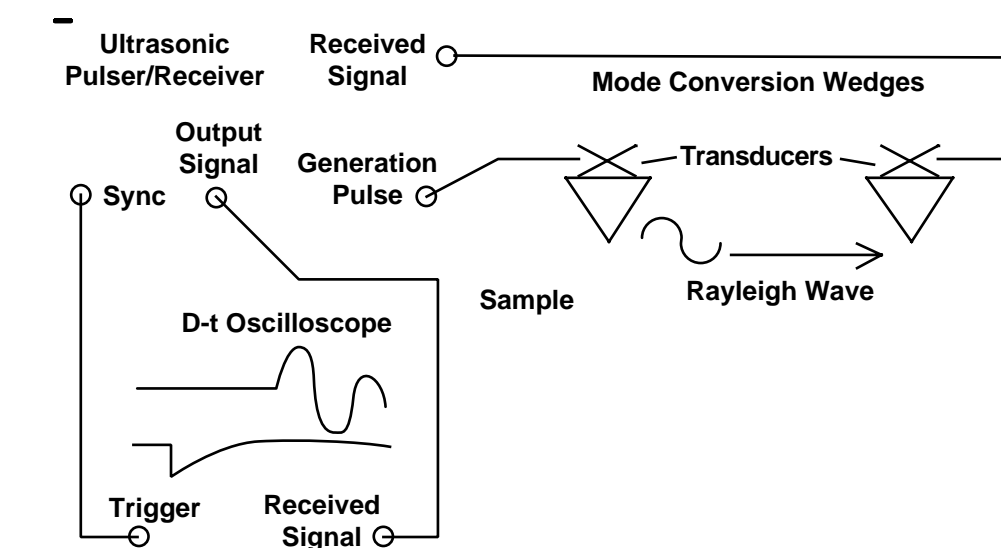


- An identical mode conversion block/transducer assembly was used for reception of the signal
- A fixed path length was achieved by mounting the mode conversion blocks on a rail
- Repeatable coupling pressure (against the sample surface) was achieved by placing a fixed mass on top of the mode conversion blocks
- Measurement of the velocity of the Rayleigh wave signal is performed upon an averaged signal of 200 waveforms

lpm 4/17/00

NDE-6

## Measurement apparatus

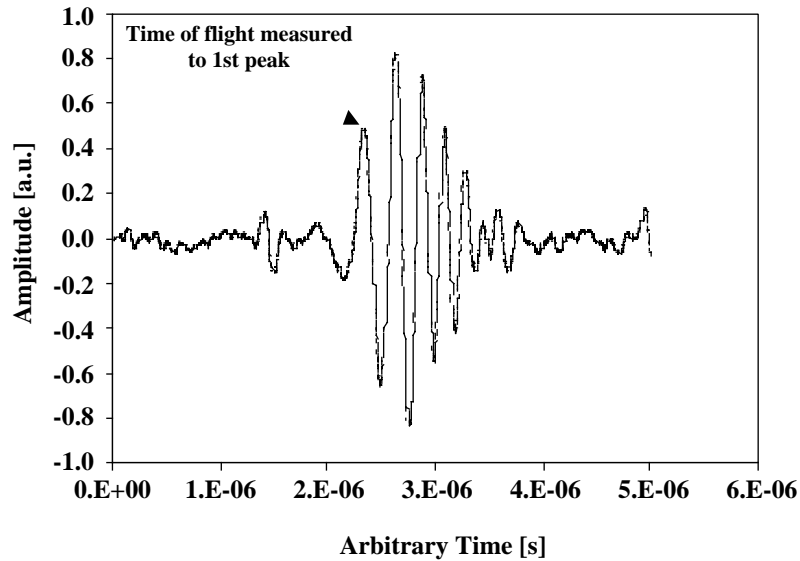


lpm 4/17/00

NDE-7



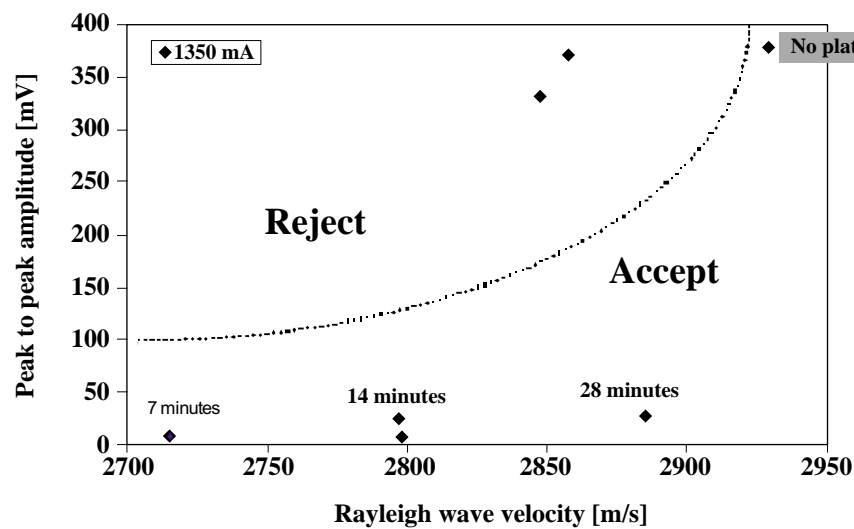
**Time of flight and peak to peak amplitude were measured from the received Rayleigh wave signals**



lpm 4/17/00

NDE-8

**Rayleigh wave velocity and amplitude can be used to define accept/reject criterion**

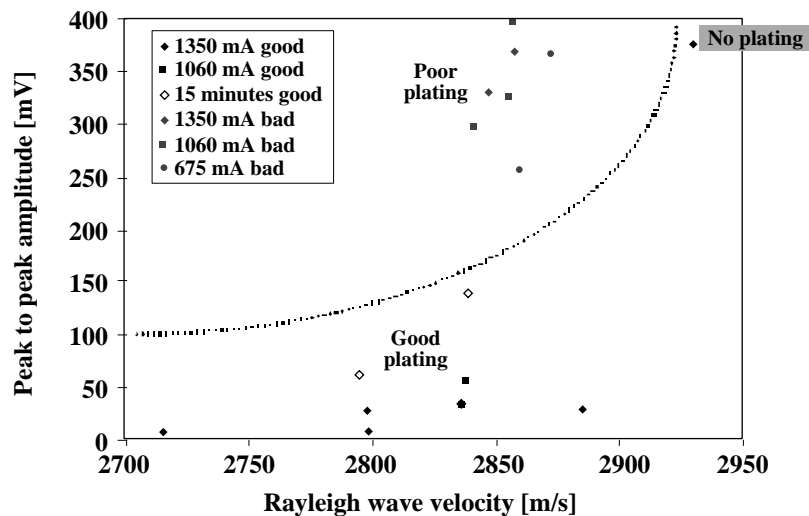


lpm 4/17/00

NDE-9



**The criterion is applicable to samples prepared using a broad range of process parameters**



lpm 4/17/00

NDE-10

**Conclusion: The Rayleigh wave technique is a viable method for on-line characterization of LHE plating**



- Rayleigh waves can be generated non-destructively in the materials of interest
- The propagation characteristics of the Rayleigh waves are strongly affected by the plating
- The Rayleigh wave velocity is dominated by the thickness of the plating
- The peak to peak amplitude is dominated by the attenuation (i.e. porosity) in the plating

lpm 4/17/00

NDE-11



# Gamma-ray Imaging with a Segmented Germanium Detector

*D. Bekedahl, J.E. Kammeraad, G.J. Schmid, LLNL;  
J.J. Blair, BECHTEL NEVADA;  
A. Kuhn, I.Y. Lee, K. Vetter, LBNL*



